

R programming

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Matrices

- ▶ Matrices are used for many purposes in R

```
> m <- rnorm(12)
> m

[1] 0.9494222 0.8166734 0.9671824 -1.7864953 -0.5218239 -0.8029565
[7] -1.8051463 -0.8884292 -0.4819043 -1.2933338 -0.2161673 -1.4497220

> dim(m) <- c(3,4)
> m

      [,1]      [,2]      [,3]      [,4]
[1,] 0.9494222 -1.7864953 -1.8051463 -1.2933338
[2,] 0.8166734 -0.5218239 -0.8884292 -0.2161673
[3,] 0.9671824 -0.8029565 -0.4819043 -1.4497220
```

- ▶ or you can specify the dimensions with the `matrix()` function

```
> m <- rnorm(12)
> m

[1] -1.83961512 0.04244126 -0.03166508 -0.47782611 -1.32690127 -1.16243158
[7] 0.13809153 0.10379338 0.26860503 0.09375672 0.66246754 0.06701164

> m <- matrix(m,nrow=3,ncol=4,byrow=F)
> m

      [,1]      [,2]      [,3]      [,4]
[1,] -1.83961512 -0.4778261 0.1380915 0.09375672
[2,] 0.04244126 -1.3269013 0.1037934 0.66246754
[3,] -0.03166508 -1.1624316 0.2686050 0.06701164
```

Matrices

► Basic functions on matrices

- `nrow()` and `ncol()` calls numbers of rows and columns
- `t()` calls the transpose of the matrix
- `rownames()` and `colnames()` are the names of columns and rows

```
> m <- rnorm(12)
> m <- matrix(m,nrow=3,ncol=4,byrow=F)
> nrow(m)
```

```
[1] 3
```

```
> ncol(m)
```

```
[1] 4
```

```
> colnames(m) <- c("A", "B", "C", "D")
> m
```

	A	B	C	D
[1,]	0.12394796	1.3154234	0.9968355	0.4413296
[2,]	0.02655774	-0.8579061	0.8457881	0.9018971
[3,]	-0.47550780	0.3490124	0.2132153	-0.5657494

Merging Vectors

- ▶ `rbind()` and `cbind()` functions merges vectors or matrices into matrices

```
> X1 <- rnorm(12)
> X2 <- 1:12
> m <- cbind(X1,X2)
> m
```

	X1	X2
[1,]	0.7179971	1
[2,]	1.9253669	2
[3,]	-0.1676504	3
[4,]	-0.7827515	4
[5,]	1.0876736	5
[6,]	-1.5190781	6
[7,]	-0.9142900	7
[8,]	-1.6296740	8
[9,]	0.2870266	9
[10,]	1.0465450	10
[11,]	1.0162229	11
[12,]	-1.8545128	12

Merging Vectors

- ▶ Number of columns should be equal for rbind
- ▶ Likewise, number of rows should be equal for cbind

```
> data_1 <- matrix(rnorm(12),nrow=3,ncol=4,byrow=T)
> data_2 <- matrix(rnorm(16),nrow=4,ncol=4,byrow=F)
> data_new <- rbind(data_1,data_2)
> data_new
```

	[,1]	[,2]	[,3]	[,4]
[1,]	-1.6178251	-0.3233475	-0.843480201	-1.2717424
[2,]	-0.7878121	-0.9117782	-0.191396380	-0.9507478
[3,]	-1.2227275	-1.1741398	-1.108056384	-0.8678325
[4,]	0.4108297	-0.7407136	-0.002151329	-1.0023822
[5,]	2.4149035	0.2027108	-1.504017600	-0.3793893
[6,]	1.0244861	0.5286680	1.392723137	1.2701646
[7,]	0.2284962	-0.4862941	1.283637517	-0.7255216

Indexing Matrices

```
> m <- matrix(rnorm(12),nrow=3,ncol=4,byrow=F)
> m
```

	[,1]	[,2]	[,3]	[,4]
[1,]	-0.5267084	-0.4637852	1.29983795	-1.232867
[2,]	0.0389733	-2.6034900	0.00954156	-1.170770
[3,]	0.5537107	0.2021856	0.43645742	-1.361930

```
> index_row <- 1:3
> index_col <- c(1,3,4)
> m[index_row,index_col]
```

	[,1]	[,2]	[,3]
[1,]	-0.5267084	1.29983795	-1.232867
[2,]	0.0389733	0.00954156	-1.170770
[3,]	0.5537107	0.43645742	-1.361930

Indexing Matrices: Examples

- ▶ Let `m` be a matrix with 7 rows where
3 rows are from the "training set"
4 rows are from the "test set"

```
> set_m <- rep(c(1,2),c(3,4))  
> set_m <- factor(set_m,levels = 1:2  
+                 ,labels = c("training","test"))  
> m <- matrix(rnorm(21),nrow=7,ncol=3,byrow=T)  
> m[set_m=="training",]
```

	[,1]	[,2]	[,3]
[1,]	-1.1759010	0.1579945	0.8915471
[2,]	0.6968994	-0.4536412	0.6820949
[3,]	-0.5826836	-0.6340029	-0.6915423

Indexing Matrices: Examples

```
> m <- cbind(m, set_m)
```

```
> m
```

				set_m
[1,]	-1.1759010	0.15799454	0.8915471	1
[2,]	0.6968994	-0.45364116	0.6820949	1
[3,]	-0.5826836	-0.63400293	-0.6915423	1
[4,]	0.8240483	0.53835903	0.6815398	2
[5,]	0.8571331	-0.01819739	1.4327240	2
[6,]	-0.7337359	-0.15168591	-0.4017226	2
[7,]	-1.4878936	-0.72207070	-0.7874091	2

```
> colnames(m) <- c("X1", "X2", "X3", "set")
```

```
> m[set_m=="training",]
```

	X1	X2	X3	set
[1,]	-1.1759010	0.1579945	0.8915471	1
[2,]	0.6968994	-0.4536412	0.6820949	1
[3,]	-0.5826836	-0.6340029	-0.6915423	1

Lists

- ▶ A list is an ordered collection of components
- ▶ Components may be arbitrary R objects (data frames, vectors, lists, etc.)
- ▶ Single bracket notation for sublists
- ▶ Double bracket notation for individual components
- ▶ Construct using the function `list()`

```
> L1 <- list(name="Fred", wife="Mary",  
+           no.children=3, child.ages=c(4,7,9))  
> L1[[3]]  
  
[1] 3
```

Lists

```
> L1 <- list(name="Fred", wife="Mary",  
+           no.children=3, child.ages=c(4,7,9))  
> L1[[3]]  
  
[1] 3  
  
> L1[c(1,3)]  
  
$name  
[1] "Fred"  
  
$no.children  
[1] 3  
  
> L1$no.children  
  
[1] 3
```

Lists: Example

```
> data_new <- matrix(rnorm(21),nrow=7,ncol=3,byrow=T)
> set_data <- rep(c(1,2),c(3,4))
> set_data <- factor(set_data,levels = 1:2,
+                    labels = c("training","test"))
> table_set <- table(set_data)
> dim_data <- dim(data_new)
> data_info <- list(data = data_new, set = set_data,
+                  set_info = table_set, dim = dim_data)
```

Lists: Example

```
> names(data_info)
[1] "data"      "set"      "set_info" "dim"

> data_info$set_info
set_data
training      test
           3      4

> data_info$dim
[1] 7 3

> data_info$set
[1] training training training test      test      test      t
Levels: training test
```

Lists: Example

```
> data_info$data
```

	[,1]	[,2]	[,3]
[1,]	0.8714042	-0.97127687	1.2170574
[2,]	0.4545998	0.27865005	0.9688815
[3,]	1.3821152	0.13976678	-1.4253669
[4,]	-1.8148906	0.86140204	0.5326529
[5,]	-1.0845851	1.41654947	-0.4814033
[6,]	0.3269725	-0.81030249	1.2556171
[7,]	-0.5696174	0.06839643	2.2507647

If Statements

```
> x <- 3
> if(x > 4){
+   cat("x is bigger than 4")
+ } else {
+   cat("x is smaller or equal than 4")
+ }

x is smaller or equal than 4
```

If Statements

```
> x <- 3
> if(x > 4){
+   cat("x is bigger than 4")
+ } else if(x==4){
+   cat("x is equal to 4")
+ } else {
+   cat("x is smaller than 4")
+ }
```

x is smaller than 4

Loops

```
> x <- rnorm(12)
> x

[1]  0.52143281 -1.93454715 -0.02574563 -0.84363691  0.48223439 -0.365
[7]  0.08245287  0.17916581  0.52421701  0.89163523  1.72646791  1.473

> for(i in 1:length(x))
+ {
+   if(x[i] > 0){
+     x[i] <- 1
+   } else {
+     x[i] <- 0
+   }
+ }
> x

[1] 1 0 0 0 1 0 1 1 1 1 1 1
```


Vectorization

- ▶ A vectorised version of the if statement is `ifelse()` function
- ▶ This is useful if you want to perform some action on every element of a vector that satisfies some condition.
- ▶ If it is possible to write code with less loops and ifs, go for it!
- ▶ It should be much faster

```
> x <- rnorm(12)
```

```
> x
```

```
[1]  1.151379534 -0.546015765 -1.583285747 -0.002819026  0.566389678  
[6] -0.379654760  0.595551404 -1.278362654 -0.637809101 -0.056317505  
[11] -1.247872021  0.689254507
```

```
> ifelse(x > 0, 1, 0)
```

```
[1] 1 0 0 0 1 0 1 0 0 0 0 1
```

Functions

```
> ifelse_new <- function(x,break_point,
+                          true_arg,false_arg){
+   y <- x
+   for(i in 1:length(y)) {
+     if(y[i] > break_point){
+       y[i] <- 1
+     } else {
+       y[i] <- 0
+     }
+   }
+   return(y)
+ }
> x <- rnorm(12)
> ifelse(x > 0, 1, 0)

[1] 0 1 0 1 1 1 0 1 0 1 1 0

> ifelse_new(x,0,1,0)

[1] 0 1 0 1 1 1 0 1 0 1 1 0
```

The apply family

- ▶ "apply" ing functions to each element of a vector/data frame/list/array
 - ▶ apply, lapply, sapply, tapply
 - ▶ apply: only used for arrays/matrices
 - ▶ lapply(): takes any structure and gives a list of results
 - ▶ tapply(): allows you to create tables of values from subgroups defined by one or more factors

```
> apply(iris[, -5], 2, mean)
```

Sepal.Length	Sepal.Width	Petal.Length	Petal.Width
5.843333	3.057333	3.758000	1.199333

The apply family

- ▶ "apply" ing functions to each element of a vector/data frame/list/array
 - ▶ `lapply()`: takes any structure and gives a list of results

```
> L1 <- list(sample(1:10,50,replace = TRUE),  
+           rep(c(1,2),30),  
+           rep(c("training","test"),each=20))  
> lapply(L1,table)
```

```
[[1]]
```

1	2	3	4	5	6	7	8	9	10
8	4	5	6	6	3	3	7	6	2

```
[[2]]
```

1	2
30	30

```
[[3]]
```

test	training
20	20

Writing Efficient R codes

► Bad code

```
> bad_code <- function(n){  
+   x <- NULL  
+   for(i in 1:n){  
+     x[i] <- sqrt(i)  
+   }  
+   return(x)  
+ }
```

► Better code

```
> good_code <- function(n){  
+   x <- sqrt(1:n)  
+   return(x)  
+ }
```

► the runtime analysis

```
> if(!"microbenchmark" %in% rownames(installed.packages()))  
+   install.packages("microbenchmark")  
> library(microbenchmark)  
> n <- 100000  
> analysis <- microbenchmark(times= 100,good_code(n),bad_code(n))
```

Writing Efficient R codes: Some examples

- ▶ Bad code: Standardizing the vector

```
> bad_code <- function(n,lower_bound,upper_bound){  
+   x <- runif(n,lower_bound,upper_bound)  
+   maxx <- max(x)  
+   minx <- min(x)  
+   for(i in 1:n){  
+     x[i] <- (x[i] - minx)/(maxx-minx)  
+   }  
+   return(x)  
+ }
```

- ▶ R has to call "-" and "/" several times
- ▶ R has to call "[]" several times

Writing Efficient R codes: Some examples

► Better code

```
> good_code <- function(n,lower_bound,upper_bound){  
+   x <- runif(n,lower_bound,upper_bound)  
+   x <- (x - min(x))/(max(x)-min(x))  
+   return(x)  
+ }
```

► "-", "/" are only called once

► "[]" has not been called.

► The runtime analysis

```
> n <- 100000  
> analysis <- microbenchmark(times= 100,  
+                               good_code(n,0,3),  
+                               bad_code(n,0,3))
```

Writing Efficient R codes: "apply" family

- ▶ Lets simulate a list with arbitrarily different sizes of samples

```
> x <- list()
> for(i in 1:100){
+   rand_x <- sample(1:100,1)
+   x[[i]] <- sample(1:100,rand_x,replace = T)
+ }
```

- ▶ Create a table for each element, put that in a list

```
> bad_code <- function(x){
+   len_x <- length(x)
+   table_x <- list()
+   for(i in 1:len_x){
+     table_x[[i]] <- table(x[[i]])
+   }
+   return(table_x)
+ }
```


Writing Efficient R codes: "apply" family

- ▶ good code, using lapply

```
> good_code <- function(x){  
+   return(lapply(x, table))  
+ }
```

- ▶ The runtime analysis

```
> analysis <- microbenchmark(times= 100,  
+                               good_code(x), bad_code(x))
```

- ▶ It is not always the case that a code is faster than the other
- ▶ But it is definitely cleaner !!!

Writing Efficient R codes: "apply" family

- ▶ It is not always the case that an "apply" function is faster
- ▶ Always use an alternative, if requires less functions
- ▶ Method 1

```
> x = 1:100
> cs_for = function(x){
+   for(i in x){
+     if(i == 1){
+       xc = x[i]
+     } else {
+       xc = c(xc, sum(x[1:i]))
+     }
+   }
+   xc
+ }
```

Writing Efficient R codes: "apply" family

► Method 2

```
> cs_apply = function(x){  
+   sapply(x, function(x) sum(1:x))  
+ }
```

► Method 3

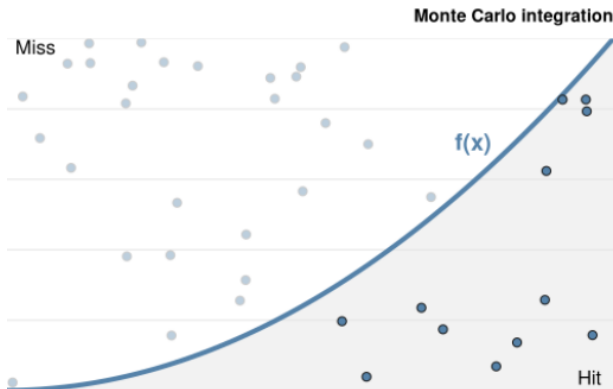
```
> analysis <- microbenchmark(cs_for(x),  
+                             cs_apply(x),  
+                             cumsum(x))
```

Writing Efficient R codes: Monte Carlo Example

- ▶ Monte Carlo Integration: Finding the Area with a collection of randomly drawn points

$$\int_0^1 x^2 dx \quad (1)$$

- ▶ Sampling random points inside a box and check if it is under the curve



Writing Efficient R codes: Monte Carlo Example

► bad code

```
> monte_carlo = function(n) {  
+   hits = 0  
+   for (i in 1:n) {  
+     u1 = runif(1)  
+     u2 = runif(1)  
+     if (u1 ^ 2 > u2)  
+       hits = hits + 1  
+   }  
+   return(hits / n)  
+ }
```

► good code

```
> monte_carlo_vec = function(n) sum(runif(n)^2 > runif(n))/n
```

► The run time analysis

```
> n <- 1000  
> analysis <- microbenchmark(monte_carlo(n), monte_carlo_vec(n))
```