

# Lab on Functional Programming and R Notebooks

## R markdown

In this lab, we will run experiments using R notebook, which is a powerful programming- and note-taking tool that allows seamless integration between code and documentation. Here is an example:

```
print("hello world!")
```

```
## [1] "hello world!"
```

You may have seen Python notebook in TB1. R notebook is exactly the same idea, but designed for R programming language. Click the “green arrow” to run your code.

Markdown is a popular scripting language that allows you to write **formatted text** in *plain text editors* (such as `notepad`) and render it later. You can switch between the source code and visualization using the “source” and “visual” in the toolbar and see the differences.

## Targets for today

In today’s lab, you will apply **functional programming** to code the pairwise distance function that we have encountered before many times. Specifically, given matrices  $A \in \mathbb{R}^{n_1 \times m}$  and  $B \in \mathbb{R}^{n_2 \times m}$ , we would like to construct a matrix  $D \in \mathbb{R}^{n_1 \times n_2}$ , where  $D_{i,j} = \text{dist}(A_{[i,\cdot]}, B_{[j,\cdot]})$ ,  $A_{[i,\cdot]}$  is  $i$ -th row of  $A$  and

$$\text{dist}(a, b) = \sqrt{\sum_i (a_i - b_i)^2}.$$

## Preparation

Before you start, I strongly recommend you review this week’s lecture slides on list and functional programming.

**Write code below**, generating two random matrices  $A$  and  $B$  with 3 rows and 2 columns which are both filled with observations from standard normal distribution (Hint: `?rnorm`).

*#type your code here and run it by clicking the green arrow on the top-right corner.*

```
A <- matrix(rnorm(3*2), 3,2)
B <- matrix(rnorm(3*2), 3,2)
```

Copy `pdist4` function from the previous lab (solution available online), and compute the distance matrix  $D$  using  $A$  and  $B$ . **NOTE:** in a previous lab we had  $B \in \mathbb{R}^{m \times n_2}$  rather than  $B \in \mathbb{R}^{n_2 \times m}$  as here. You will need to modify `pdist4` to take this into account.

```
# Copy and modify the pdist4 code here
# fully vectorised version
pdist4 <- function(A,B){
  B <- t(B) # Easy solution: traspose B upfront.
  nA <- nrow(A)
  nB <- ncol(B)
  o1 <- matrix(1, 1, nB)
```

```

o2 <- matrix(1, nA, 1)

D <- sqrt( rowSums(A^2) %*% o1 - 2 * A%*%B + o2%*%t(colSums(B^2)) )
return(D)
}

# type your code here
pdist4(A,B)

##           [,1]      [,2]      [,3]
## [1,] 0.8779723 1.435905 2.731130
## [2,] 1.4344275 1.966185 3.203766
## [3,] 1.8195973 2.390001 3.147154

# Optional checking whether our code is correct!
M <- matrix(NA, 3, 3)
for(ii in 1:nrow(A)){
  for(jj in 1:nrow(B)){
    M[ii, jj] <- sqrt(sum((A[ii, ]-B[jj, ])^2))
  }
}
M - pdist4(A,B) # Looks right!

##           [,1]      [,2]      [,3]
## [1,]      0 2.220446e-16 -4.440892e-16
## [2,]      0 2.220446e-16 0.000000e+00
## [3,]      0 0.000000e+00 -4.440892e-16

```

## Pairwise Distance

We already know how to write vectorized code for computing  $D$ . Now let us inspect at this computation from a functional programming point of view.

- Given two data matrices  $A \in \mathbb{R}^{n_1 \times m}$  and  $B \in \mathbb{R}^{n_2 \times m}$ ,
- Fix  $k$ , and compute  $(A_{[i,k]} - B_{[j,k]})^2$  for all  $i, j$ .
- Store the outcome as the  $i, j$ -th element of a matrix  $D_k$ .
- Apply the above computation for all  $k$ .
- $D := \sqrt{\sum_k D_k}$

### Task 1

Write a function `diff2` that takes two scalars  $a, b$ , and compute  $(a - b)^2$ :

```

#type your code here
diff2 <- function(a, b){
  return( (a-b)^2 )
}

```

`outer(X,Y,FUN)` applies function `FUN` to all pairs of elements in `X` and `Y`. In other words, it computes a matrix  $D$ ,  $D_{i,j} = \text{fun}(X_i, Y_j), \forall i, j$ . For example,

```

a <- c(1,2,3,4)
b <- c(1,2,3,4)

a_plus_b <- function(a,b){

```

```

    return (a+b)
}

# function as an input variable
# outer applies the operation a_plus_b to each pair of elements in a and b
outer(a, b, a_plus_b)

```

```

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

```

*# what is the  $i, j$ -th element in the matrix below?*

Now use `outer` and `diff2`, compute a 3 by 3 matrix  $D^{(1)}$ ,  $D_{i,j}^{(1)} = \text{diff2}(A_{[i,1]}, B_{[j,1]})$ .

*#type your code here*

```

D1 <- outer(A[,1], B[,1], diff2)
D1

##      [,1]      [,2]      [,3]
## [1,] 0.1630108 0.6927171 1.8120319
## [2,] 0.2391202 0.8418971 1.5897783
## [3,] 1.1350245 2.2318139 0.4685237

```

## Task 2

We would like to repeat the above operation for all  $k$ . We can use a for loop but, in functional programming, we tend to think we “apply” a data operation to **the entire dataset** rather than create loops that iterate over each piece of our dataset.

This is exactly what we did above: We applied `diff2` to all pairs of  $i, j$ .

Let us create another function `D_k`, that takes one integer input  $k$  and compute  $D_k$  using `outer` (wrap the code you have just written in a function)

```

# type your code here
D_k <- function(k){
  return(outer(A[,k], B[,k], diff2))
}

```

Now apply this function to a list of numbers (hint: `?lapply`). The list contains integers from 1 to  $m$ , where  $m$  is 2 in this toy case.

```

# type your code here
res <- lapply(list(1,2), D_k)

```

You should get a list of matrices whose  $k$ -th element is  $D_k$ .

## Task 3

Now, all that is left is summing  $D_k$  together and we are done! Can we use a for loop to sum all  $D_k$  matrices together?

In Functional Programming, we tend to think our program is a huge pipeline that applies functions repeatedly to our data: `op3(op2(op1(data)))...` (recall the “data pipeline” in the lecture slides)

We can rewrite the summation of  $D_k$  using this paradigm:

$\sum_{k=1 \dots 4} D_k$  is the same as

`sum2(sum2(sum2(D_1,D_2),D_3),D_4)`, where `sum2(X,Y)` sums two matrices X and Y together. Write a function `sum2` that sums over two matrices

```
# type your code here
```

```
sum2 <- function(a,b){  
  return(a+b)  
}
```

Now we need to apply this function to the list of  $D_k$  recursively. This can be done using a function called `Reduce`. `Reduce(f,x)` applies `f` to a list `x` recursively. In other words, it computes

`f(f(f(x_1, x_2),x_3),x_4)...`

```
# type your code here
```

```
res2 <- Reduce(sum2, res)
```

Square root the outcome.

```
# type your code here
```

```
D <- sqrt(res2)  
D
```

```
##           [,1]      [,2]      [,3]  
## [1,] 0.8779723 1.435905 2.731130  
## [2,] 1.4344275 1.966185 3.203766  
## [3,] 1.8195973 2.390001 3.147154
```

Check, is this the same result you obtained using `pdist4`?

```
# Your checking code here
```

```
D - pdist4(A,B)
```

```
##           [,1]      [,2]      [,3]  
## [1,] 0 2.220446e-16 -4.440892e-16  
## [2,] 0 2.220446e-16 0.000000e+00  
## [3,] 0 0.000000e+00 -4.440892e-16
```

#### Task 4

`lapply(l, fun)` applies a function `fun` to a list `l`.

There is a sibling function called `apply(A, MARGIN, fun)` which applies the function `fun` to a matrix along a certain margin. See `?apply`.

Setting `MARGIN` to 1 means you are applying `fun` to each row of the matrix.

Setting `MARGIN` to 2 means you are applying `func` to each column of the matrix.

`apply` applies `fun` to rows/columns of `A`, return a vector containing the results.

```
X <- matrix(c(1,2,3,4), 2, 2)
X
```

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

```
# compute the average of a vector
average <-function(r) {
  return(sum(r)/length(r))
}
```

```
# compute row average
apply(X, 1, average)
```

```
## [1] 2 3
```

Now, given the D matrix you obtained. Could you sort the elements of each row of D, in ascending order?

Hint: `sort` function, takes in a vector, and sort them according to the ascending order. `?sort`

```
sort(c(4,3,2,1))
```

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

```
# your code here
```

```
D
```

```
##      [,1]      [,2]      [,3]
## [1,] 0.8779723 1.435905 2.731130
## [2,] 1.4344275 1.966185 3.203766
## [3,] 1.8195973 2.390001 3.147154
```

```
t(apply(D, 1, sort))
```

```
##      [,1]      [,2]      [,3]
## [1,] 0.8779723 1.435905 2.731130
## [2,] 1.4344275 1.966185 3.203766
## [3,] 1.8195973 2.390001 3.147154
```