

# Programming for Data Science

Apply family in R language

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# Apply family in R

- How to efficiently apply a function to each element of array, data frame and list.

*For instance: to apply a function to the rows/columns of a matrix*

- Functions `apply`, `lapply`, `sapply`, `tapply` can be used:

`apply` : only used for arrays/matrices;

`lapply` : takes any data structure and gives a list of results;

`sapply` : like `lapply`, but it tries to simplify the result to a vector or matrix if possible;

`tapply` : allows us to apply a function on a subset of values grouped according to one or more factors.

## Function apply()

- the apply function returns a vector or array of values obtained by applying a function to margins of an array or matrix.

```
apply(X, MARGIN, FUN, ...)
```

**X** : array;

**MARGIN** : 1 for rows, 2 for columns;

**FUN** : one function to be applied;

**...** : optional arguments to **FUN**;

```
> m
```

```
      [,1]      [,2]
[1,] -0.1767643 -0.1950407
[2,]  1.5306045  0.3307676
[3,] -0.3806768  0.8992097
```

```
> apply(m, 1, sum)  by rows
```

```
[1] -0.3718050 1.8613721 0.5185329
```

```
> apply(m, 2, sum) by columns
```

```
[1] 0.9731634 1.0349366
```

## Function apply()

- the apply function returns a vector or array of values obtained by applying a function to margins of an array or matrix.

```
apply(X, MARGIN, FUN, ...)
```

**X** : array;

**MARGIN** : 1 for rows, 2 for columns;

**FUN** : one function to be applied;

**...** : optional arguments to **FUN**;

```
> m
```

	[, 1]	[, 2]
[1,]	-0.1767643	-0.1950407
[2,]	1.5306045	0.3307676
[3,]	-0.3806768	0.8992097

```
> apply(m, 1, max) by rows
```

```
[1] -0.1767643 1.5306045 0.8992097
```

```
> apply(m, 2, min) by columns
```

```
[1] -0.3806768 -0.1950407
```

# Function apply()

- the apply function returns a vector or array of values obtained by applying a function to margins of an array or matrix.

```
apply(X, MARGIN, FUN, ...)
```

**X** : array;

**MARGIN** : 1 for rows, 2 for columns;

**FUN** : one function to be applied;

**...** : optional arguments to **FUN**;

```
> m = matrix(rnorm(6), nrow = 2)
```

	[, 1]	[, 2]
[1,]	0.34963183	1.0360705
[2,]	-1.04686386	0.6846824
[3,]	-0.06385193	0.6219289

```
> apply(m, 1, quantile, prob = c(0.25, 0.75))
```

**by columns**

	[, 1]	[, 2]	[, 3]
25%	0.5212415	-0.6139773	0.1075933
75%	0.8644608	0.2517958	0.4504837

# Function lapply()

- the lapply function returns a list where each element is the result of applying a function to the corresponding element of input data structure.

`lapply(X, FUN, ...)`

`X` : any data that can be compatible with a list;

`FUN` : one function to be applied;

`...` : optional arguments to `FUN`;

```
> data()      to list in-built data set
> lapply(trees, mean)  trees is a in-built data set
$Girth
[1]13.24839
$Height
[1]76
$Volume
[1]30.17097
```

```
> trees
  Girth Height Volume
1   8.3    70   10.3
2   8.6    65   10.3
3   8.8    63   10.2
4  10.5    72   16.4
5  10.7    81   18.8
6  10.8    83   19.7
7  11.0    66   15.6
8  11.0    75   18.2
9  11.1    80   22.6
10 11.2    75   19.9
```

# Function `sapply()`

- the `sapply` function is a user-friendly version and wrapper of "`lapply`" by default returning a vector, matrix .

`sapply(X, FUN, ...)`

- `X` : any data that can be compatible with a list/vector/matrix;
- `FUN` : one function to be applied;
- `...` : optional arguments to `FUN`;

`> data()` to list in-built data set

`> sapply(trees, mean)` trees is a in-built data set

<i><b>Girth</b></i>	<i><b>Height</b></i>	<i><b>Volume</b></i>
13.24839	76.00000	30.17097

```
> trees
  Girth Height Volume
1   8.3    70   10.3
2   8.6    65   10.3
3   8.8    63   10.2
4  10.5    72   16.4
5  10.7    81   18.8
6  10.8    83   19.7
7  11.0    66   15.6
8  11.0    75   18.2
9  11.1    80   22.6
10 11.2    75   19.9
```

# Function tapply()

- the tapply function allows us to apply a function on a subset of values grouped according to one or more factors .

tapply(X, INDEX, FUN, ...)

- X** : any data that can be compatible with a list;
- INDEX** : list of one or more factors used to cluster X;
- FUN** : one function to be applied;
- ...** : optional arguments to **FUN**;

> library(MASS)      to load MASS data set

> Cars93      Car93 is a MASS data set

	Manufacturer	Model	Type	Min.Price	Price	Max.Price	MPG.city
1	Acura	Integra	Small	12.9	15.9	18.8	25
2	Acura	Legend	Midsize	29.2	33.9	38.7	18
3	Audi	90	Compact	25.9	29.1	32.3	20
4	Audi	100	Midsize	30.8	37.7	44.6	19
5	BMW	535i	Midsize	23.7	30.0	36.2	22
6	Buick	Century	Midsize	14.2	15.7	17.3	22
7	Buick	LeSabre	Large	19.9	20.8	21.7	19
8	Buick	Roadmaster	Large	22.6	23.7	24.9	16
9	Buick	Riviera	Midsize	26.3	26.3	26.3	19

> tapply(Cars93\$Price, Cars93\$Manufacturer, mean)

for each brand

Compute the average price



# How to apply a function in parallel

- In R different possibilities exist (e.g. `parlapply`, `mclapply`, `clusterApply`, ...), but not all are portable (i.e. they can not be used indifferently on Window, Linux and macOS);
- **“snow”** package provides parallelization functionality on Window, Linux and macOS;

```
> install.packages("snow")
```

- It supports: Socket and Message Passing Interface (MPI) protocols;

## Socket

- ▶ Portable, but low level protocol;
- ▶ Can be used interactively;
- ▶ Good for running on a multicore machine;

## MPI

- ▶ Needs the **“Rmpi”** package;
- ▶ Cannot be used interactively;
- ▶ Good for running on several nodes;
- ▶ Works everywhere where **Rmpi** is installed.

## How to apply a function in parallel

BASE R	SNOW
<code>lapply</code>	<code>parLapply</code>
<code>sapply</code>	<code>parSapply</code>
<code>apply(rowwise)</code>	<code>parRapply, parApply(,1)</code>
<code>apply(columnwise)</code>	<code>parCapply, parApply(,2)</code>

- To use these function you have to initialize the cluster with command `makeCluster()`
- When computation is terminated you have to close the cluster with command `stopCluster()`

# How to apply a function in parallel

- An example using 6 cores and `Sys.sleep()`

```
> z = list(1, 1, 1, 1, 1, 1)
```

```
> cl <- makeCluster(6)
```

```
> system.time(parSapply(cl, z, Sys.sleep))
```

it is ~ 6 times faster than `sapply`

```
> stopCluster(cl)
```

# How to apply a function in parallel

- An example using 8 cores and [an own function](#)

```
> z = rexp(6000000, 6)
```

```
> z = list(z, z, z, z, z, z, z, z)
```

```
> cl <- makeCluster(8)
```

```
> system.time(parSapply(cl, z, function(val){2^val + 2^val + 2^val}))  
it is ~ 2 times faster than sapply
```

```
> stopCluster(cl)
```

# How to apply a function in parallel

- An example using 8 cores and [an own function](#)

```
> z = rexp(6000000, 6)
```

```
> z = list(z, z, z, z, z, z, z, z)
```

```
> cl <- makeCluster(8)
```

```
> system.time(parSapply(cl, z, function(val){val + val}))  
it is ~ 3 times slower than sapply
```

```
> stopCluster(cl)
```

# Parallelization Efficiency

- The time spent in each invocation of the worker function should not be too short;
- If the time spent in each invocation of the worker function vary very much, try the load balancing versions of the functions (e.g. `clusterApplyLB`);
- Avoid copying large things back and forth:
  - ▶ Export large datasets up front with `clusterExport()`;
  - ▶ Write the worker function to return as little as possible.

# Parallelization Efficiency

- In general, it is recommend using forking instead of sockets if you are not on Windows;

- Forking with `parSapply` implemented in library `parallel`

```
> library(parallel)
```

```
> z = rexp(6000000, 6)
```

```
> z = list(z, z, z, z, z, z, z, z)
```

```
> cl <- makeForkCluster(nnodes = 8)
```

```
> system.time(parSapply(cl, z, function(val){2^val + 2^val + 2^val}))  
it is ~ 3 times faster than supply
```

```
> stopCluster(cl)
```

# Function `do.call()`

- it constructs and executes a function call from a name or a function and a list of arguments to be passed to it..

`do.call(what, args)`

`what` : character string naming the function to be called;

`args` : a list of arguments to the function call. The names attribute of `args` gives the argument names;

```
> a = c(2,2,2,2)
```

```
> f = c("mean", "sum")
```

```
> do.call(f[1], list(a))
```

```
[1] 2
```

```
> do.call(f[2], list(a))
```

```
[1] 8
```



## Exercises on apply

- Compute sums of the columns of the hills data set (in library MASS);
- Compute row and column sums of a matrix 10x10 whose values are generated according to uniform distribution between 4 and 10;
- Use apply to calculate the standard deviation of the columns of a matrix;
- Create a list of vectors of varying length (using sample() function);
- Consider in-built data set "airquality" compute the average wind speed and ozone percentage with respect to "month" column.

# Exercises on apply

- Compute sums of the columns of the hills data set(in library MASS);
  - > *lapply(hills, sum)*
  - > *sapply(hills, sum)*

## Exercises on apply

- Compute row and column sums of a matrix 10x10 whose values are generated according to uniform distribution between 4 and 10

```
> m = matrix(runif(100, min = 4, max = 10), ncol = 10)
> apply(m, 1, sum)
> apply(m, 2, sum)
```

# Exercises on apply

- Use apply to calculate the standard deviation of the columns of a matrix.

```
> m = matrix(runif(100, min = 4, max = 10), ncol = 10)
```

```
> apply(m, 2, sd)
```

# Exercises on apply

- Create a list of vectors of varying length (using `sample()` function)

```
> veclen = sample(11 : 40)  
> mylist = lapply(veclen, runif)
```

or

```
> mylist = lapply(sample(11 : 40, 10), runif)
```

## Exercises on apply

- Consider in-built data set "airquality" compute the average wind speed and ozone percentage with respect to "month" column.

```
> tapply(airquality$Wind, airquality$Month, mean)
```

```
> tapply(airquality$Ozone, airquality$Month, mean, na.rm = TRUE)
```

**na.rm=TRUE removes NA from mean computation**

# Exercises on apply

- Compute a hundred times the mean of 1000000 observations distributed as  $\mathcal{N}(0, 1)$  using `lapply` and `parLapply`. Compare the execution times using `rbenchmark` package.

## Exercises on apply

- Compute a hundred times the mean of 1000000 observations distributed as  $\mathcal{N}(0, 1)$  using `lapply` and `parLapply`. Compare the execution times using `rbenchmark` package.

```
> install.packages("rbenchmark")
```

```
> library(rbenchmark)
```

```
> cl <- makeCluster(4)
```

```
> benchmark(  
  parLapply(cl, 1 : 100, function(x){mean(rnorm(1000000))}),  
  lapply(1 : 100, function(x){mean(rnorm(1000000))}), replications = 5)
```

```
> stopCluster(cl)
```

```
test replications  
2      lapply(1:100, function(x) {\n  mean(rnorm(1e+06))\n})      5  
1 parLapply(cl, 1:100, function(x) {\n  mean(rnorm(1e+06))\n})      5  
elapsed relative user.self sys.self user.child sys.child  
2 32.650  2.933  32.607  0.000  0  0  
1 11.132  1.000  0.013  0.009  0  0
```