# R language and data analysis: apply family

Qiang Shen

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#### vectorization

- ▶ R user vs. R programmer/developer
- Vectorization is the more limited process of converting a computer program from a scalar implementation, which processes a single pair of operands at a time, to a vector implementation which processes one operation on multiple pairs of operands at once.
- Vectorization is a particular form of how parallelism is achieved.

#### vectorization

- apply family
- plyr/reshape
- dplyr/data.table

## Looping on the Command Line

- apply: Apply a function over the margins of an array
- lapply: Loop over a list and evaluate a function on each element
- sapply: Same as lapply but try to simplify the result
- tapply: Apply a function over subsets of a vector
- mapply: Multivariate version of lapply

split is also useful, especially in conjunction with lapply.

- what does apply mean.
- apply to different dimension.
- compare with rowMeans etc.
- different functions.

apply is used to evaluate a function (often an anonymous one) over the **margins** of an array.

- used to apply a function to the rows or columns of a matrix or an array.
- used with general arrays, e.g. taking the average, standard deviation.
- ▶ not really faster than writing a loop, but make it simpler.

## apply: margin

#### **Bivariate Probability Distribution**

Example - Two discrete rv's X and Y

Bivariate pdf							
	%	0	1	Pr(X)			
X	0	1/8	0	1/8			
	1	2/8	1/8	3/8			
	2	1/8	2/8	3/8			
	3	0	1/8	1/8			
	Pr(Y)	4/8	4/8	1			

Figure 1:

```
str(apply)
```

```
## function (X, MARGIN, FUN, ...)
```

- X is an array
- MARGIN is an integer vector indicating which margins should be "retained".
- FUN is a function to be applied
- ... is for other arguments to be passed to FUN

apply a function to the rows or columns of a matrix.



Figure 2: r

# apply vs. standard method

apply(x, 2, sum)

```
y<-matrix(rnorm(6),2,3)
У
cbind(mean(y[1,]), mean(y[2,]))
apply(y,1,mean)
x <- matrix(rnorm(30), 5, 6)
sumx<-NULL
for (i in 1:6){
temp < -sum(x[,i])
sumx[i]<-temp</pre>
sumx
```

## col/row sums and means

- ► For sums and means of matrix dimensions, we have some shortcuts.
- rowSums(x)
- rowMeans(x)
- colSums(x)
- colMeans(x)

apply with ...

```
x <- matrix(rnorm(200), 20, 10)
dim(x)
apply(x, 2, quantile)
apply(x, 2, quantile, probs = c(0.25, 0.75))</pre>
```

## lapply :start from an example

lapply Loop over a list and evaluate a function on each element

```
##user-defined function.
  func<-function(x){
    if (x\%2 == 0) {
      ret<-'even'
    }else{
      ret<-'odd'}
    return(ret)
func(101)
vec<-round(runif(4)*100)
vec
vec;func(vec)
lapply(vec,func)
```

vectorization.

```
func<-Vectorize(func)
func(vec)
# ifelse(vec%%2, 'even', 'odd')</pre>
```

## lapply

```
lapply takes three arguments:
```

```
str(lapply)
## function (X, FUN, ...)
```

lapply: beyond apply.

lapply always returns a list, regardless of the input.

lapply(iris[,1:4],mean)

## lapply

```
x <- list(a = 1:5, b = rnorm(10))
x;lapply(x, mean)</pre>
```

## lapply ...

```
x < -1:4
lapply(x, runif, min = 0, max = 10)
## [[1]]
## [1] 2.1
##
## [[2]]
## [1] 6.7 6.7
##
## [[3]]
## [1] 9.0 3.3 2.0
##
## [[4]]
## [1] 0.56 5.85 1.45 6.85
```

### lapply

make use of *anonymous* functions for lapply. example: An anonymous function for extracting the 1st row of each matrix.

```
data <- list(a = matrix(1:6, 2, 3), b = matrix(1:6, 3, 2),
data
lapply(data, function(x) x[1,])</pre>
```

### lapply

anonymous functions continued.

```
lapply(iris[,1:4],function(x) sd(x,na.rm=T)/mean(x,na.rm=T)
myfunc<-function(x){
  rec<-c(mean(x,na.rm=T),sd(x,na.rm=T))
  return(rec)
}
result<-lapply(iris[,1:4],myfunc)
result</pre>
```

methods to covert list into data.frame.

```
t(as.data.frame(result))
# t(sapply(result, '['))
do.call('rbind',result)
```

## sapply

sapply will try to simplify the result of lapply if possible.

- ▶ If the result is a list where every element is length 1, then a vector is returned
- ▶ If the result is a list where every element is a vector of the same length (> 1), a matrix is returned.
- ▶ If it can't figure things out, a list is returned

# tapply: split-apply-combine.

- Split up a big dataset
- Apply a function to each piece
- Combine all the pieces back together
- map-reduce in hadoop.

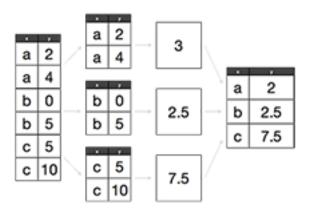


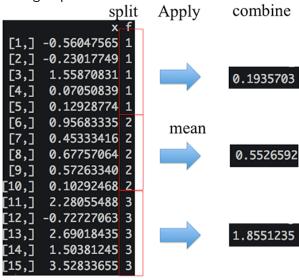
Figure 3:

 ${\tt tapply}$  is used to apply a function over subsets of a vector. - apply, sapply/lapply vs.  ${\tt tapply}$ 

```
str(tapply)
```

```
## function (X, INDEX, FUN = NULL, ..., default = NA, simp
```

Take group means.



```
set.seed(123)
x \leftarrow c(rnorm(5), runif(5), rnorm(5, 1))
f < -gl(3, 5)
data<-cbind(x,f)
tapply(x, f, mean)
## 1 2 3
## 0.19 0.55 1.86
tapply(x, f, mean, simplify = FALSE)
```

```
tapply(x, f, mean, simplify = FALSE)
## $`1`
```

```
## [1] 0.19
##
## $`2`
```

## [1] 0.55

## ## \$`3`

Find group ranges.

```
## $`1`
## [1] -0.56 1.56
##
## $`2`
## [1] 0.10 0.96
##
## $`3`
## [1] -0.73 3.53
```

tapply(x, f, range)

- iris flower dataset:Stata,python,R
- https://en.wikipedia.org/wiki/Iris\_flower\_data\_set



Figure 4: iris

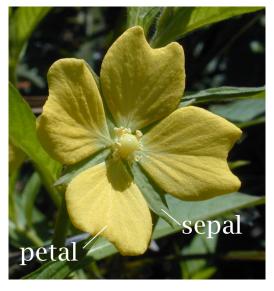


Figure 5: iris

str(iris)

vs. pivot table in excel

```
'data.frame': 150 obs. of 5 variables:
   $ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4
##
```

\$ Sepal.Width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 ##

## \$ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1

\$ Petal.Width : num 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0 ## ## \$ Species : Factor w/ 3 levels "setosa", "versicolog

## head(iris)

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Spe
## 1	5.1	3.5	1.4	0.2	set

## 2	4.9	3.0	1.4	0.2 se
## 3	4.7	3.2	1.3	0.2 se
## 4	4.6	3.1	1.5	0.2 se

vs. pivot table in excel

```
tapply(iris[,1], iris$Species, mean)
##
      setosa versicolor virginica
                            6.6
##
         5.0
              5.9
t(sapply(iris[,1:4],function(x) tapply(x, iris$Species, meaning)
##
              setosa versicolor virginica
## Sepal.Length 5.01
                          5.9
                                   6.6
## Sepal.Width 3.43 2.8 3.0
## Petal.Length 1.46 4.3 5.6
## Petal.Width 0.25
                     1.3 2.0
# myfun < -function(x) {
# tapply(x, iris\$Species, mean, na.rm=T)
# }
```

### split

split takes a vector or other objects and splits it into groups determined by a factor or list of factors.

```
## function (x, f, drop = FALSE, ...)
# function (x, f, drop = FALSE, ...)
```

- x is a vector (or list) or data frame
- ▶ f is a factor (or coerced to one) or a list of factors
- drop indicates whether empty factors levels should be dropped

### split

##

[1]

```
x \leftarrow c(rnorm(10), runif(10), rnorm(10, 1))
f \leftarrow gl(3, 10)
split(x, f)
## $`1`
   [1] 0.549 0.238 -1.049 1.295 0.826 -0.056 -0.784 -0
##
##
## $ 2
    [1] 0.139 0.233 0.466 0.266 0.858 0.046 0.442 0.799 0.3
##
## $`3`
```

## split

## ## \$`3` ## [1] 0.53

A common idiom is split followed by an lapply.

```
## $`1`
## [1] -0.027
##
## $`2`
## [1] 0.39
```

lapply(split(x, f), mean)

# Splitting on More than One Level

```
x \leftarrow rnorm(10)
f1 \leftarrow g1(2, 5)
f2 \leftarrow g1(5, 2)
f1;f2
## [1] 1 1 1 1 1 2 2 2 2 2
## Levels: 1 2
## [1] 1 1 2 2 3 3 4 4 5 5
## Levels: 1 2 3 4 5
interaction(f1, f2)
```

```
## [1] 1.1 1.1 1.2 1.2 1.3 2.3 2.4 2.4 2.5 2.5
## Levels: 1.1 2.1 1.2 2.2 1.3 2.3 1.4 2.4 1.5 2.5
```

```
data<-cbind(x,f1,f2)
data</pre>
```

# split:Empty levels can be dropped.

```
## List of 6
## $ 1.1: num [1:2] -0.772 0.287
## $ 1.2: num [1:2] -1.221 0.435
## $ 1.3: num 0.8
## $ 2.3: num -0.164
## $ 2.4: num [1:2] 1.243 -0.934
## $ 2.5: num [1:2] 0.394 0.404
```

## Splitting a Data Frame

##

```
s <- split(iris, iris$Species)</pre>
# str(s)
sapply(s, function(x) colMeans(x[, 1:4],na.rm=T))
```

```
setosa versicolor virginica
## Sepal.Length 5.01
                          6.6
                     5.9
## Sepal.Width 3.43 2.8 3.0
## Petal.Length 1.46 4.3 5.6
## Petal.Width 0.25 1.3 2.0
```

```
# sapply(s, function(x) sapply((x[, 1:4]), mean, na.rm=T))
```

```
lapply and sapply only iterate over a single R object.

str(lapply)

## function (X, FUN, ...)

str(mapply)
```

## function (FUN, ..., MoreArgs = NULL, SIMPLIFY = TRUE, US

lapply and sapply only iterate over a single R object.

```
## list(rep(1, 4), rep(2, 3), rep(3, 2), rep(4, 1))
mapply(rep,1:4,4:1)
```

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

lapply and sapply only iterate over a single R object.

```
## list(rep(1, 4), rep(2, 3), rep(3, 2), rep(4, 1))
mapply(rep,1:4,4:1)
```

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

```
noise <- function(n, mean, sd) {</pre>
rnorm(n, mean, sd)
}
noise(5,1,2)
s<-noise(1:100, 1:20, 2)
S
mean(s)
sd(s)
##revisit this when we talk about Sprintf function later.
simulation < -mapply(noise, 100000, 1:20, 5:1)
dim(simulation)
apply(simulation, 2, mean)
apply(simulation,2,sd)
```

```
##1. generate dataset called data
data < -data.frame(a < -c(1:5), b < -c(2:6), d < -c(3:7).
                  e<-c('1','2','a','x','y'),
                  f < -c('2', '3', '5', 'd', 'c').
                  g < -c('3', 'k', '5', '6', NA),
                  stringsAsFactors = F)
data[,1:3] <-sapply(data[,1:3], as.character)
names(data)<-letters[1:6]
##2. do mapply
myfunc<-function(x, y) {</pre>
  ifelse(grepl('[a-z]',v), v, x)
data2<-data
data2[1:3] = mapply(myfunc, data[1:3], data[4:6])
data2
```

## Looping on the Command Line

- ► lapply: Loop over a list and evaluate a function on each element
- sapply: Same as lapply but try to simplify the result
- apply: Apply a function over the margins of an array
- tapply: Apply a function over subsets of a vector
- mapply: Multivariate version of lapply