talk08 练习与作业

目录

0.1 练习和作业说明

将相关代码填写入以 “‘{r} “‘ 标志的代码框中，运行并看到正确的结果；

完成后，用工具栏里的”Knit” 按键生成 PDF 文档；

将 PDF 文档改为：姓名`-`学号`-talk08` 作业`.pdf`，并提交到老师指定的平台/钉群。

0.2 talk08 内容回顾

- for loop
- `apply` functions

- dplyr 的本质是遍历
- map functions in purrr package
- 遍历与并行计算

0.3 练习与作业：用户验证

请运行以下命令，验证你的用户名。

如你当前用户名不能体现你的真实姓名，请改为拼音后再运行本作业！

```
Sys.info()[["user"]]
```

```
## [1] "lucas"
```

```
Sys.getenv("HOME")
```

```
## [1] "/Users/lucas"
```

0.4 练习与作业 1：loop 初步

0.4.1 loop 练习（部分内容来自 r-exercises.com 网站）

1. 写一个循环，计算从 1 到 7 的平方并打印 print；
2. 取 iris 的列名，计算每个列名的长度，并打印为下面的格式：Sepal.Length (12)；
3. 写一个 while 循环，每次用 rnorm 取一个随机数字并打印，直到取到的数字大于 1；
4. 写一个循环，计算 Fibonacci 序列的值超过 1 百万所需的循环数；注：Fibonacci 序列的规则为：0, 1, 1, 2, 3, 5, 8, 13, 21 ...；

```r
## 代码写这里，并运行；
# Task 01
for (i in 1:7) {
  result <- i^2
  print(result)
}
```

```
## [1] 1
## [1] 4
## [1] 9
## [1] 16
## [1] 25
## [1] 36
## [1] 49
```

```r
# Task 02
column_names =
  names(iris)
for (col_name in column_names) {
  col_name_length =
    nchar(col_name)
  cat(col_name,
      "(",
      col_name_length,
      ")\n")
}
```

```
## Sepal.Length ( 12 )
## Sepal.Width ( 11 )
## Petal.Length ( 12 )
## Petal.Width ( 11 )
## Species ( 7 )
```

```r
# Task 03
random_number =
  rnorm(1)
while (random_number <= 1) {
  print(random_number)
  random_number =
    rnorm(1)
}

# Task 04
fibonacci =
  c(0, 1)
count = 2
while (
  tail(fibonacci, 1) +
  fibonacci[length(fibonacci) - 1]
  <= 1000000) {
  next_fib =
    tail(fibonacci, 1) +
    fibonacci[length(fibonacci) - 1]
  fibonacci =
    c(fibonacci, next_fib)
  count =
    count + 1
}
print(count)
```

```
## [1] 31
```

0.5 练习与作业 2：loop 进阶，系统和其它函数

0.5.1 生成一个数字 matrix，并做练习

生成一个 100 x 100 的数字 matrix:

1. 行、列平均，用 rowMeans, colMeans 函数;
2. 行、列平均，用 apply 函数
3. 行、列总和，用 rowSums, colSums 函数;
4. 行、列总和，用 apply 函数
5. 使用自定义函数，同时计算:
    - 行平均、总和、sd
    - 列平均、总和、sd

```
## 代码写这里，并运行;
# Prepare the matrix
set.seed(123)
mat =
  matrix(
    rnorm(10000),
    nrow = 100)

# Task 01
row_means =
  rowMeans(mat)
col_means =
  colMeans(mat)

# Task 02
# Average number of rows
row_means_apply =
  apply(mat, 1, mean)
# Average number of cols
col_means_apply =
  apply(mat, 2, mean)
```

```r
# Task 03
row_sums =
  rowSums(mat)
col_sums =
  colSums(mat)

# Task 04
# Sum number of rows
row_sums_apply =
  apply(mat, 1, sum)
# Sum number of cols
col_sums_apply =
  apply(mat, 2, sum)

# Task 05
custom_stats =
  function(x) {
  return(
    c(
      mean =
        mean(x),
      sum =
        sum(x),
      sd =
        sd(x)))
}

# Calculating row statistics
row_stats =
  t(apply(mat, 1, custom_stats))

# Calculating col statistics
col_stats =
```

```
  apply(mat, 2, custom_stats)

# Print the results
print("Row Means:")
```

```
## [1] "Row Means:"
```

```
head(row_means, n = 3)
```

```
## [1] -0.02668423 -0.08311430 -0.18342356
```

```
print("Col Means:")
```

```
## [1] "Col Means:"
```

```
head(col_means, n = 3)
```

```
## [1]  0.09040591 -0.10754680  0.12046511
```

```
print("Row Means (apply):")
```

```
## [1] "Row Means (apply):"
```

```
head(row_means_apply, n = 3)
```

```
## [1] -0.02668423 -0.08311430 -0.18342356
```

```
print("Col Means (apply):")
```

```
## [1] "Col Means (apply):"
```

```r
head(col_means_apply, n = 3)
```

```
## [1]  0.09040591 -0.10754680  0.12046511
```

```r
print("Row Sums:")
```

```
## [1] "Row Sums:"
```

```r
head(row_sums, n = 3)
```

```
## [1]  -2.668423  -8.311430 -18.342356
```

```r
print("Col Sums:")
```

```
## [1] "Col Sums:"
```

```r
head(col_sums, n = 3)
```

```
## [1]   9.040591 -10.754680  12.046511
```

```r
print("Row Sums (apply):")
```

```
## [1] "Row Sums (apply):"
```

```r
head(row_sums_apply, n = 3)
```

```
## [1]  -2.668423  -8.311430 -18.342356
```

```r
print("Col Sums (apply):")
```

```
## [1] "Col Sums (apply):"
```

```r
head(col_sums_apply,n = 3)
```

```
## [1]   9.040591 -10.754680  12.046511
```

```r
print("Custom Row Statistics:")
```

```
## [1] "Custom Row Statistics:"
```

```r
head(row_stats,n = 3)
```

```
##             mean        sum        sd
## [1,] -0.02668423  -2.668423 0.9589688
## [2,] -0.08311430  -8.311430 0.9371213
## [3,] -0.18342356 -18.342356 1.0357881
```

```r
print("Custom Col Statistics:")
```

```
## [1] "Custom Col Statistics:"
```

```r
head(col_stats,n = 3)
```

```
##            [,1]        [,2]       [,3]        [,4]       [,5]        [,6]
## mean 0.09040591  -0.1075468  0.1204651 -0.03622291  0.1058509 -0.04229996
## sum  9.04059086 -10.7546798 12.0465110 -3.62229084 10.5850925 -4.22999578
## sd   0.91281588   0.9669866  0.9498790  1.03878122  0.9893458  0.93872815
##             [,7]       [,8]       [,9]       [,10]      [,11]       [,12]
## mean  -0.1496441  0.1058735 0.09358971 -0.01919274  0.1199406 -0.01917556
## sum  -14.9644141 10.5873547 9.35897144 -1.91927403 11.9940576 -1.91755583
## sd     1.0282366  1.0100100 1.05180659  1.02033166  1.0429982  0.99767147
##             [,13]       [,14]      [,15]      [,16]      [,17]      [,18]
## mean -0.002539253 -0.06021128 0.09298927 0.03548695 0.08903406 0.03105473
## sum  -0.253925262 -6.02112781 9.29892666 3.54869516 8.90340573 3.10547347
## sd    0.928007043  1.01382919 0.98614286 0.88929622 1.09648466 1.02443405
```

```
##           [,19]      [,20]       [,21]       [,22]       [,23]      [,24]
## mean 0.06518418 0.07288886  -0.1184286  -0.1328738  -0.1038760 0.07303035
## sum  6.51841753 7.28888607 -11.8428579 -13.2873799 -10.3876010 7.30303453
## sd   1.04354899 1.08710170   0.8867420   0.8671375   0.9234582 1.01178395
##           [,25]      [,26]       [,27]       [,28]      [,29]      [,30]
## mean -0.03805934  0.1060876  -0.1149985 -0.01519487 0.04754731 0.09564058
## sum  -3.80593362 10.6087594 -11.4998513 -1.51948694 4.75473105 9.56405766
## sd    0.99456268  0.9751193   1.0359941  0.94454890 1.01726182 1.10530160
##           [,31]       [,32]      [,33]      [,34]       [,35]      [,36]
## mean -0.05389207  -0.1543210 0.03879604 0.05114384 0.005222176 0.06795743
## sum  -5.38920679 -15.4321040 3.87960435 5.11438357 0.522217582 6.79574304
## sd    1.07991180   0.8705622 1.07626589 0.84481290 1.047729024 1.07836057
##           [,37]       [,38]      [,39]     [,40]      [,41]       [,42]
## mean -0.01763709  -0.2499180 0.07249396  0.148551 -0.0830015 -0.04217516
## sum  -1.76370864 -24.9917976 7.24939600 14.855096 -8.3001504 -4.21751621
## sd    0.97329530   0.8768267 0.91526278  1.095703  1.1023174  0.78330286
##           [,43]       [,44]       [,45]      [,46]      [,47]       [,48]
## mean -0.01227263 -0.04126713 -0.05545621 0.06173919 0.05972319 -0.06701344
## sum  -1.22726277 -4.12671273 -5.54562141 6.17391864 5.97231916 -6.70134424
## sd    1.01742000  1.08455734  1.00191901 1.00882076 0.96523497  0.98094922
##           [,49]       [,50]      [,51]      [,52]      [,53]       [,54]
## mean -0.05332069 -0.08863744 0.06803094 0.03556477 -0.0782075 -0.03531996
## sum  -5.33206943 -8.86374425 6.80309354 3.55647721 -7.8207500 -3.53199625
## sd    0.94231314  1.11142003 0.88940618 0.86387437  0.9641162  1.00713013
##           [,55]      [,56]       [,57]      [,58]      [,59]      [,60]
## mean  0.1002368 0.08108622 -0.02297392 0.01378047 0.03658164  0.1460111
## sum  10.0236830 8.10862250 -2.29739242 1.37804736 3.65816433 14.6011054
## sd    1.0771408 1.01709562  1.01292952 0.98644219 1.02524416  1.0675179
##           [,61]       [,62]      [,63]       [,64]      [,65]      [,66]
## mean 0.05149166  -0.2379236 0.08115196  -0.1689926  0.1111330  0.1544749
## sum  5.14916559 -23.7923602 8.11519559 -16.8992567 11.1133042 15.4474925
## sd   0.93443052   1.1540099 1.04262975   1.0431884  0.9165161  0.9676328
##           [,67]       [,68]     [,69]       [,70]      [,71]       [,72]
```

```
## mean -0.09230807  -0.1635888 0.0468654  -0.1113391 0.05067045 -0.04929186
## sum  -9.23080748 -16.3588773 4.6865402 -11.1339105 5.06704482 -4.92918617
## sd    0.87696277   1.0257226 0.9645486   0.9469142 1.01823607  0.87801873
##           [,73]        [,74]      [,75]       [,76]      [,77]       [,78]
## mean  0.1629300 -0.005236446  0.1073088  -0.1465294 0.09553372 -0.03913617
## sum  16.2929984 -0.523644574 10.7308791 -14.6529425 9.55337199 -3.91361747
## sd    0.8941483  1.008655635  0.9944261   1.0769782 0.96638316  1.01031030
##           [,79]       [,80]      [,81]       [,82]      [,83]     [,84]
## mean 0.05489841  -0.1765393 -0.0459872 -0.00540858  0.1283985  0.154827
## sum  5.48984112 -17.6539282 -4.5987198 -0.54085797 12.8398513 15.482699
## sd   1.05630156   1.0659721  1.0161120  1.07501409  0.9959257  0.969051
##           [,85]      [,86]       [,87]      [,88]      [,89]       [,90]
## mean 0.01586283 0.08128843 -0.04852987 0.02417753  0.1095373  -0.2148035
## sum  1.58628257 8.12884279 -4.85298714 2.41775337 10.9537307 -21.4803458
## sd   1.11207550 1.06293318  0.94756707 1.04856646  1.0506158   1.1474328
##            [,91]       [,92]       [,93]       [,94]      [,95]       [,96]
## mean 0.006974302 -0.06152943 -0.05830342 -0.09212069 0.05559785 -0.06251836
## sum  0.697430160 -6.15294266 -5.83034172 -9.21206910 5.55978494 -6.25183599
## sd   0.994143844  0.98719987  1.03118035  0.83965976 1.06511882  0.98403060
##           [,97]       [,98]       [,99]      [,100]
## mean -0.0103759 -0.09106338  -0.1305609 -0.03451664
## sum  -1.0375903 -9.10633772 -13.0560897 -3.45166380
## sd    0.9922376  0.91410182   1.0321389  0.99787848
```

0.5.2 用 mtcars 进行练习

用 tapply 练习:

1. 用 **汽缸数**分组，计算 **油耗**的 **平均值**;
2. 用 **汽缸数**分组，计算 **wt** 的 **平均值**;

用 dplyr 的函数实现上述计算

```
## 代码写这里，并运行;
# Using tapply
# Task 01
cylinder_mpg =
  tapply(
    mtcars$mpg,
    mtcars$cyl,
    mean)
print(cylinder_mpg)
```

```
##        4        6        8
## 26.66364 19.74286 15.10000
```

```
# Task 02
cylinder_wt =
  tapply(
    mtcars$wt,
    mtcars$cyl,
    mean)
print(cylinder_wt)
```

```
##        4        6        8
## 2.285727 3.117143 3.999214
```

```
# using dplyr
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##     filter, lag
```

```
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union
```

```
# Task 01
cylinder_mpg_dplyr =
  mtcars %>%
  group_by(cyl) %>%
  summarise(
    mean_mpg =
      mean(mpg))

print(cylinder_mpg_dplyr)
```

```
## # A tibble: 3 x 2
##     cyl mean_mpg
##   <dbl>    <dbl>
## 1     4     26.7
## 2     6     19.7
## 3     8     15.1
```

```
# Task 02
cylinder_wt_dplyr =
  mtcars %>%
  group_by(cyl) %>%
  summarise(
    mean_wt =
      mean(wt))

print(cylinder_wt_dplyr)
```

```
## # A tibble: 3 x 2
##     cyl mean_wt
##   <dbl>   <dbl>
```

```
## 1     4    2.29
## 2     6    3.12
## 3     8    4.00
```

0.5.3 练习 lapply 和 sapply

1. 分别用 lapply 和 sapply 计算下面 list 里每个成员 vector 的长度:

```
list( a = 1:10, b = letters[1:5], c = LETTERS[1:8] );
```

2. 分别用 lapply 和 sapply 计算 mtcars 每列的平均值;

```
## 代码写这里，并运行;
# Task 01
my_list =
  list(a = 1:10, b = letters[1:5], c = LETTERS[1:8])

# Using lapply
lengths_lapply =
  lapply(my_list, length)

# Using sapply
lengths_sapply =
  sapply(my_list, length)

# Print the result
print("Using lapply:")
```

```
## [1] "Using lapply:"
```

```r
print(lengths_lapply)
```

```
## $a
## [1] 10
##
## $b
## [1] 5
##
## $c
## [1] 8
```

```r
print("Using sapply:")
```

```
## [1] "Using sapply:"
```

```r
print(lengths_sapply)
```

```
##  a  b  c
## 10  5  8
```

```r
# Task 02
# Using lapply
avg_by_column_lapply =
  lapply(mtcars, mean)
print("Using lapply:")
```

```
## [1] "Using lapply:"
```

```r
print(avg_by_column_lapply)
```

```
## $mpg
## [1] 20.09062
##
```

```
## $cyl
## [1] 6.1875
##
## $disp
## [1] 230.7219
##
## $hp
## [1] 146.6875
##
## $drat
## [1] 3.596563
##
## $wt
## [1] 3.21725
##
## $qsec
## [1] 17.84875
##
## $vs
## [1] 0.4375
##
## $am
## [1] 0.40625
##
## $gear
## [1] 3.6875
##
## $carb
## [1] 2.8125
```

```
# Using sapply
avg_by_column_sapply =
  sapply(mtcars, mean)
print("Using sapply:")
```

```
## [1] "Using sapply:"
```

```
print(avg_by_column_sapply)
```

```
##        mpg        cyl       disp         hp       drat         wt       qsec
##  20.090625   6.187500 230.721875 146.687500   3.596563   3.217250  17.848750
##         vs         am       gear       carb
##   0.437500   0.406250   3.687500   2.812500
```

0.6 练习与作业 3：loop 进阶，purr 包的函数

0.6.1 map 初步

生成一个变量：

```
df <- tibble(
  a = rnorm(10),
  b = rnorm(10),
  c = rnorm(10),
  d = rnorm(10)
)
```

用 map 计算：

- 列平均值、总和和中值

```
## 代码写这里，并运行；
# Load the packages
library(purrr)
```

```r
library(dplyr)

# Preparing the data
df = tibble(
  a = rnorm(10),
  b = rnorm(10),
  c = rnorm(10),
  d = rnorm(10)
)

# Calculating
results =
  map(df,
      ~c(mean = mean(.),
         sum = sum(.),
         median = median(.)))

# Bind the results
result_df =
  bind_rows(results)

# Print the results
print(result_df)
```

```
## # A tibble: 4 x 3
##     mean   sum  median
##    <dbl> <dbl>   <dbl>
## 1  0.317  3.17  0.0291
## 2 -0.278 -2.78 -0.153
## 3  0.225  2.25 -0.0483
## 4  0.177  1.77 -0.0715
```

0.6.2 map 进阶

用 map 配合 purr 包中其它函数，用 mtcars：

为每一个 **汽缸数**计算燃油效率 mpg 与重量 wt 的相关性（Pearson correlation），得到 p 值和 correlation coefficient 值。

```
## 代码写这里，并运行;
# Load the package
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ------------------------ tidyverse 2.0.0 --
## v forcats   1.0.0     v stringr   1.5.0
## v ggplot2   3.4.3     v tibble    3.2.1
## v lubridate 1.9.2     v tidyr     1.3.0
## v readr     2.1.4
## -- Conflicts ------------------------------------------ tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts
```

```
# Calculating
result =
  mtcars %>%
  group_by(cyl) %>%
  nest() %>%
  mutate(
    correlation =
      map(data, ~cor.test(.$mpg, .$wt)),
    p_value =
      map_dbl(correlation, "p.value"),
    correlation_coefficient =
      map_dbl(correlation, "estimate"))
```

```r
# Print the result
print(result$p_value)
```

```
## [1] 0.09175766 0.01374278 0.01179281
```

```r
print(result$correlation_coefficient)
```

```
## [1] -0.6815498 -0.7131848 -0.6503580
```

0.6.3 keep 和 discard

1. 保留 iris 中有 factor 的列，并打印前 10 行；
2. 去掉 iris 中有 factor 的列，并打印前 10 行；

```r
## 代码写这里，并运行；
library(dplyr)

# Task 01
iris_with_factors =
  iris %>% keep(is.factor)
head(iris_with_factors, 10)
```

```
##    Species
## 1   setosa
## 2   setosa
## 3   setosa
## 4   setosa
## 5   setosa
## 6   setosa
## 7   setosa
## 8   setosa
```

```
## 9   setosa
## 10  setosa
```

```
# Task 02
iris_without_factors =
  iris %>% discard(is.factor)
head(iris_without_factors, 10)
```

```
##    Sepal.Length Sepal.Width Petal.Length Petal.Width
## 1           5.1         3.5          1.4         0.2
## 2           4.9         3.0          1.4         0.2
## 3           4.7         3.2          1.3         0.2
## 4           4.6         3.1          1.5         0.2
## 5           5.0         3.6          1.4         0.2
## 6           5.4         3.9          1.7         0.4
## 7           4.6         3.4          1.4         0.3
## 8           5.0         3.4          1.5         0.2
## 9           4.4         2.9          1.4         0.2
## 10          4.9         3.1          1.5         0.1
```

0.6.4 用 reduce

用 reduce 得到以下三个 vector 中共有的数字:

```
c(1, 3, 5, 6, 10),
  c(1, 2, 3, 7, 8, 10),
  c(1, 2, 3, 4, 8, 9, 10)
```

```
## 代码写这里，并运行;
# Load the package
library(purrr)
```

```r
# Create three vectors
vector1 =
  c(1, 3, 5, 6, 10)
vector2 =
  c(1, 2, 3, 7, 8, 10)
vector3 =
  c(1, 2, 3, 4, 8, 9, 10)

# Find the commom_elements
common_elements =
  reduce(
    list(
      vector1,
      vector2,
      vector3),
    intersect)

# Print the results
print(common_elements)
```

```
## [1]  1  3 10
```

0.6.5 运行以下代码，观察得到的结果，并用 tidyverse 包中的 spread 等函数实现类似的结果

```r
dfs <- list(
  age = tibble(name = "John", age = 30),
  sex = tibble(name = c("John", "Mary"), sex = c("M", "F")),
  trt = tibble(name = "Mary", treatment = "A")
);

dfs %>% reduce(full_join);
```

```
## 代码写这里，并运行;
# Example
library(purrr)
library(tidyverse)
dfs <- list(
  age = tibble(name = "John", age = 30),
  sex = tibble(name = c("John", "Mary"), sex = c("M", "F")),
  trt = tibble(name = "Mary", treatment = "A")
);

dfs_join =
  dfs %>% reduce(full_join);
```

```
## Joining with `by = join_by(name)`
## Joining with `by = join_by(name)`
```

```
# Task
# Load the library
library(tidyverse)

# Create data frame
data_full =
  data.frame(
    `Name` = character(),
    `Key` = character(),
    `Value` = character()
  )

data_list =
  list(
    data.frame(
      Name = "John",
      Key = "Age",
```

```
      Value = "30"
    ),
    data.frame(
      Name = "John",
      Key = "Sex",
      Value = "M"
    ),
    data.frame(
      Name = "Mary",
      Key = "Sex",
      Value = "F"
    ),
    data.frame(
      Name = "John",
      Key = "Treatment",
      Value = "A"
    )
  )

data_full =
  do.call(rbind, data_list)


# Spread it
result_tidyverse =
  data_full %>%
  spread(key = `Key`, value = `Value`)

result_tidyverse$Age = as.numeric(result_tidyverse$Age)

print(result_tidyverse)
```

```
##   Name Age Sex Treatment
```

```
## 1 John  30   M         A
## 2 Mary  NA   F      <NA>
```

0.7 练习与作业 4：pmap 和 map 的更多用法

请参考 https://r4ds.had.co.nz/iteration.html 的 Mapping over multiple arguments 部分

0.7.1 map2

运行以下代码，查看输出结果。用 for 循环重现计算结果。

```
mu <- list(5, 10, -3);
sigma <- list(1, 5, 10);
map2(mu, sigma, rnorm, n = 5)
```

```
## 代码写这里，并运行;
# Example
mu <- list(5, 10, -3);
sigma <- list(1, 5, 10);
map2(mu, sigma, rnorm, n = 5)
```

```
## [[1]]
## [1] 5.749278 4.463402 4.366117 6.065397 6.322708
##
## [[2]]
## [1] 11.091503  3.712076 14.126324  6.264098  9.179697
##
## [[3]]
## [1]  10.63365394 -19.43866101   4.45335944  -0.06955236  -0.29811515
```

```r
# Task
mu = list(5, 10, -3)
sigma = list(1, 5, 10)
n = 5

# Caculating
results = vector("list", length(mu))

# Using 'for'
for (i in 1:length(mu)) {
  results[[i]] =
    rnorm(
      n,
      mean = mu[[i]],
      sd = sigma[[i]])
}

# Print the result
print(results)
```

```
## [[1]]
## [1] 5.262239 4.697671 5.152479 4.662372 5.247955
##
## [[2]]
## [1]  6.914691 11.571684 17.371821  5.284800  7.628093
##
## [[3]]
## [1] -8.632784  8.689311 -6.331444  6.852970 -3.547983
```

0.7.2 pmap

运行以下代码，查看输出结果。用 for 循环重现计算结果。

```r
params <- tribble(
```

```
  ~mean, ~sd, ~n,
    5,     1,  1,
   10,     5,  3,
   -3,    10,  5
)
params %>%
  pmap(rnorm)
```

```
## 代码写这里，并运行;
# Example
params <- tribble(
  ~mean, ~sd, ~n,
    5,     1,  1,
   10,     5,  3,
   -3,    10,  5
)
params %>%
  pmap(rnorm)
```

```
## [[1]]
## [1] 4.357922
##
## [[2]]
## [1] 14.166952  6.769464  7.615705
##
## [[3]]
## [1]  -7.742218   6.559735  18.120634 -16.470769 -11.629351
```

```
print(params)
```

```
## # A tibble: 3 x 3
##    mean    sd     n
##   <dbl> <dbl> <dbl>
```

```
## 1     5     1     1
## 2    10     5     3
## 3    -3    10     5
```

```r
# Task
# Load the library
library(dplyr)

# Processing
params = tribble(
  ~mean, ~sd, ~n,
    5,     1,  1,
   10,     5,  3,
   -3,    10,  5
)

result_list = list()

for (i in 1:nrow(params)) {
  mean_val = params$mean[i]
  sd_val = params$sd[i]
  n_val = params$n[i]

  result_list[[i]] =
    c(mean_val, sd_val, n_val)
}

result_df =
  as.data.frame(do.call(rbind, result_list))
colnames(result_df) =
  c("mean", "sd", "n")

print(result_df)
```

```
##   mean sd n
## 1    5  1 1
## 2   10  5 3
## 3   -3 10 5
```

0.8 练习与作业 5：并行计算

0.8.1 安装相关包，成功运行以下代码，观察得到的结果，并回答问题

* parallel
* foreach
* iterators

```
library(parallel);
library(foreach);
```

```
##
## Attaching package: 'foreach'

## The following objects are masked from 'package:purrr':
##
##     accumulate, when
```

```
library(iterators);

## 检测有多少个 CPU --
( cpus <- parallel::detectCores() );
```

```
## [1] 8
```

```
## 创建一个 data.frame
d <- data.frame(x=1:10000, y=rnorm(10000));

## make a cluster --
cl <- makeCluster( cpus - 1 );

## 分配任务 ...
res <- foreach( row = iter( d, by = "row" ) ) %dopar% {
  return ( row$x * row$y  );
}
```

```
## Warning: executing %dopar% sequentially: no parallel backend registered
```

```
## 注意在最后关闭创建的 cluster
stopCluster( cl );

summary(unlist(res));
```

```
##      Min.   1st Qu.    Median      Mean   3rd Qu.      Max.
## -31531.17  -2702.29    -18.66   -103.84   2411.58  32933.75
```

问：你的系统有多少个 CPU？此次任务使用了多少个？**答**：用代码打印出相应的数字即可：

```
## 代码写这里，并运行；
cpus = parallel::detectCores()
cpus_used = cpus - 1

# Print the data
print(cpus)
```

```
## [1] 8
```

```
print(cpus_used)
```

```
## [1] 7
```

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