## StudyR2

PX

3/8/2020

## Group Manipulation

```
# apply only works for Matrix, meaning all of the elements must be same type,
# either character, numeric, logical.
# If used on other object like data.frame, it will be converted into matrix first.
# apply(arg1, arg2, arg3)
# arg1: object to work on
# arg2: margin to apply function over
# 1 operate over rows, 2 operate over columns
# arg3: function we want to apply
babyM <- matrix(1:9, nrow = 3)</pre>
apply(babyM,1,sum)
## [1] 12 15 18
apply(babyM,2,sum)
## [1] 6 15 24
babyM[2,1] \leftarrow NA
apply(babyM, 1, sum)
## [1] 12 NA 18
apply(babyM, 1, sum, na.rm = TRUE)
## [1] 12 13 18
# lapply apply a function to each element of a list and return results as a list
babyList <- list(A = matrix(1:9, 3), B =1:5, c = matrix(1:6, 2), D=2)
babyList
## $A
        [,1] [,2] [,3]
##
## [1,]
          1
## [2,]
           2
                     8
## [3,]
##
## $B
## [1] 1 2 3 4 5
##
## $c
```

```
[,1] [,2] [,3]
##
## [1,]
        1 3 5
## [2,]
          2
               4 6
##
## $D
## [1] 2
lapply(babyList, sum)
## $A
## [1] 45
##
## $B
## [1] 15
##
## $c
## [1] 21
##
## $D
## [1] 2
# mapply
# apply a function to each element of multiple lists.
firstlist <- list(A = matrix(1:12,4), B = matrix(1:16,2), C = 1:7)
secondlist \leftarrow list(A = matrix(1:12,4), B = matrix(1:16,4), c =25:15)
mapply(identical,firstlist,secondlist)
##
       Α
             В
## TRUE FALSE FALSE
simF <- function(x,y){</pre>
 nrow(x)+nrow(y)
mapply(simF, firstlist, secondlist)
## $A
## [1] 8
##
## $B
## [1] 6
##
## $C
## integer(0)
# Aggregate: sum
# formula consists of a left side and right side separated by tilde(~)
# Left side represents a variable that we want to calculate
# Right side represents one or more variables that we want to group the calculation by
require(ggplot2)
```

## Loading required package: ggplot2

```
data("diamonds")
# price should be broken up by cut
aggregate( price ~ cut, diamonds, mean)
##
           cut
                  price
## 1
          Fair 4358.758
## 2
          Good 3928.864
## 3 Very Good 3981.760
## 4
      Premium 4584.258
## 5
         Ideal 3457.542
# price grouped by cut and color
aggregate(price ~ cut + color, diamonds, mean, na.rm=TRUE)
##
            cut color
                          price
## 1
           Fair
                     D 4291.061
## 2
                    D 3405.382
           Good
## 3
     Very Good
                     D 3470.467
## 4
        Premium
                    D 3631.293
## 5
          Ideal
                    D 2629.095
## 6
           Fair
                    E 3682.312
## 7
           Good
                    E 3423.644
## 8
                    E 3214.652
     Very Good
## 9
        Premium
                    E 3538.914
## 10
          Ideal
                    E 2597.550
## 11
                    F 3827.003
           Fair
## 12
           {\tt Good}
                    F 3495.750
## 13 Very Good
                    F 3778.820
## 14
        Premium
                    F 4324.890
                    F 3374.939
## 15
          Ideal
                    G 4239.255
## 16
           Fair
## 17
           Good
                    G 4123.482
## 18 Very Good
                     G 3872.754
## 19
                     G 4500.742
        Premium
## 20
          Ideal
                     G 3720.706
## 21
           Fair
                    H 5135.683
## 22
           Good
                    H 4276.255
## 23 Very Good
                    H 4535.390
## 24
        Premium
                    H 5216.707
## 25
          Ideal
                    H 3889.335
## 26
           Fair
                    I 4685.446
                    I 5078.533
## 27
           Good
                    I 5255.880
## 28 Very Good
## 29
        Premium
                    I 5946.181
## 30
          Ideal
                    I 4451.970
## 31
                     J 4975.655
           Fair
## 32
           Good
                     J 4574.173
## 33 Very Good
                     J 5103.513
## 34
        Premium
                     J 6294.592
## 35
          Ideal
                     J 4918.186
# group two variable, use cbind()
aggregate(cbind(price, carat)~cut, diamonds, mean, na.rm=TRUE)
```

```
##
         cut
                 price
## 1
         Fair 4358.758 1.0461366
## 2
         Good 3928.864 0.8491847
## 3 Very Good 3981.760 0.8063814
## 4 Premium 4584.258 0.8919549
## 5
        Ideal 3457.542 0.7028370
PLY-R
# plyr package "split-apply-combine"
        input: data.frame output:data.frame
# ldply
           input: list output: data.frame
# llply
           input: list output: list
require(plyr)
## Loading required package: plyr
data("baseball")
head(baseball)
##
             id year stint team lg g ab r h X2b X3b hr rbi sb cs bb so ibb
## 4
      ansonca01 1871
                       1 RC1
                                  25 120 29 39 11
                                                    3 0
                                                         16
                                                              6
                                                                2
                                                                   2
## 44 forceda01 1871
                        1 WS3
                                  32 162 45 45
                                                9
                                                    4 0
                                                          29 8 0 4 0
                                                                         NA
## 68 mathebo01 1871
                        1 FW1
                                  19 89 15 24
                                                   1 0 10 2 1 2 0
                        1 NY2
                                  33 161 35 58 5 1 1 34 4 2 3 0 NA
## 99 startjo01 1871
## 102 suttoez01 1871
                        1 CL1
                                  29 128 35 45
                                               3
                                                   7 3
                                                          23
                                                             3 1 1 0
                       1 CL1
## 106 whitede01 1871
                                  29 146 40 47 6 5 1 21 2 2 4 1 NA
      hbp sh sf gidp
##
       NA NA NA
## 4
## 44
       NA NA NA
## 68
       NA NA NA
                 NA
## 99
       NA NA NA
                NA
## 102 NA NA NA
                 NA
## 106 NA NA NA
                 NA
# set year before 1954, sf all zeros
baseball$sf[baseball$year < 1954] <- 0
# check if it works
any(is.na(baseball$sf))
## [1] FALSE
# set NA hbp to O
baseball$hbp[is.na(baseball$hbp)] <- 0</pre>
# check
any(is.na(baseball$hbp))
## [1] FALSE
# only keep players with at least 50 at bats in a season
baseball <- baseball[baseball$ab >= 50,]
# calcualte OBP
```

baseball\$OBP <- with(baseball, (h+bb+hbp)/(ab+bb+hbp+sf))</pre>

```
tail(baseball)
               id year stint team lg
                                      g ab r
                                                h X2b X3b hr rbi sb cs
                                                                         bb
## 89499 claytro01 2007
                        1 TOR AL 69 189 23 48
                                                          0 1
                                                    14
                                                               12
                                                                   2 1
                                                                         14
                                                          2 2
## 89502 cirilje01 2007
                          1 MIN AL 50 153 18 40
                                                      9
                                                               21
                                                                    2 0
                                                                         15
                                                                             13
## 89521 bondsba01 2007
                          1 SFN NL 126 340 75 94
                                                     14
                                                          0 28
                                                               66
                                                                   5 0 132
                                                                             54
                          1 HOU NL 141 517 68 130
## 89523 biggicr01 2007
                                                     31
                                                          3 10
                                                               50
                                                                   4
                                                                      3
                                                                         23 112
## 89530 ausmubr01 2007
                           1 HOU NL 117 349 38 82
                                                    16
                                                          3 3
                                                               25
                                                                   6 1
                                                                         37
                                                                             74
## 89533 aloumo01 2007
                           1 NYN NL 87 328 51 112 19
                                                         1 13 49 3 0
                                                                             30
        ibb hbp sh sf gidp
                                 OBP
             1 3 3
## 89499
         0
                         8 0.3043478
## 89502
         0
             1 3 2
                         9 0.3274854
              3 0 2 13 0.4800839
## 89521 43
## 89523 0
              3 7 5
                        5 0.2846715
              6 4 1
## 89530
          3
                        11 0.3180662
              2 0 3
## 89533
          5
                        13 0.3916667
# with function, this allows us to sepcify the columns of a data.frame
# without having to specify the data.frame name each time
# build a function
obp <- function(data)</pre>
 c(OBP = with(data, sum(h+bb+hbp)/sum(ab+bb+hbp+sf)))
# use ddply to calculate career OBP for each player
careerOBP <- ddply(baseball, .variables = "id", .fun = obp)</pre>
# sort the result by OBPs
careerOBP <- careerOBP[order(careerOBP$OBP,decreasing = TRUE), ]</pre>
head(careerOBP,10)
##
              id
                       OBP
## 1089 willite01 0.4816861
## 875
        ruthba01 0.4742209
## 658
       mcgrajo01 0.4657478
## 356
       gehrilo01 0.4477848
       bondsba01 0.4444622
## 85
## 476
       hornsro01 0.4339068
        cobbty01 0.4329655
## 184
## 327
        foxxji01 0.4290509
## 953 speaktr01 0.4283386
## 191 collied01 0.4251246
# llply
babyList
## $A
        [,1] [,2] [,3]
##
## [1,]
          1
               4
                    7
## [2,]
          2
                    8
## [3,]
          3
               6
                    9
```

```
##
## $B
## [1] 1 2 3 4 5
##
## $c
##
        [,1] [,2] [,3]
## [1,]
          1
               3
## [2,]
           2
                4
                     6
##
## $D
## [1] 2
llply(babyList,sum)
## $A
## [1] 45
##
## $B
## [1] 15
##
## $c
## [1] 21
## $D
## [1] 2
# useful functions in plyr
# each, apply multiple functions to a function
aggregate(price ~ cut, diamonds, each(mean, median))
           cut price.mean price.median
## 1
                 4358.758
                              3282.000
          Fair
## 2
                              3050.500
          Good
                 3928.864
## 3 Very Good
                              2648.000
                 3981.760
                              3185.000
## 4
      Premium
                 4584.258
## 5
         Ideal
                 3457.542
                              1810.000
# idata.frame, creates a reference to a data.frame
# so that subsetting is much faster and more memory efficient.
system.time(dlply(baseball, "id", nrow))
##
      user system elapsed
##
     0.072
            0.000 0.072
iBaseball <-idata.frame(baseball)</pre>
system.time(dlply(iBaseball, "id", nrow))-system.time(dlply(baseball, "id", nrow))
##
      user system elapsed
     0.016 0.003 0.019
Data Table
# using plyr can be very slow.
# data.table extends data.frame
```

```
# data.table have an index like databases
# fast access values, group by operation and joins
require(data.table)
## Loading required package: data.table
theDT <- data.table(A = 1:10, B = letters[1:10], C = LETTERS[11:20], D = rep(c("One", "Two", "Three"),1
theDT
##
       ABC
                 D
## 1: 1 a K
## 2: 2 b L
               Two
## 3: 3 c M Three
## 4: 4 d N
              One
## 5: 5 e O
              Two
## 6: 6 f P Three
## 7: 7 g Q
               One
## 8: 8 h R
               Two
## 9: 9 i S Three
## 10: 10 j T
               One
class(theDT$B)
## [1] "character"
# by default data.frame turns character data into factors
# while data.table does not
theDT[theDT$A <=3, ]</pre>
     A B C
## 1: 1 a K
             One
## 2: 2 b L
## 3: 3 c M Three
# the big difference between data.table and data.frame
# to access columns
# column should be specified as a list instead of a character vector
theDT[, list(A,D, C)]
             D C
##
       Α
## 1: 1
           One K
## 2: 2
           Two L
## 3: 3 Three M
## 4: 4
           One N
## 5: 5
           Two O
## 6: 6 Three P
## 7: 7
           One Q
## 8: 8
           Two R
## 9: 9 Three S
## 10: 10
           One T
theDT[, A]
## [1] 1 2 3 4 5 6 7 8 9 10
```

```
theDT[,list(A)]
##
       Α
## 1: 1
## 2: 2
## 3: 3
## 4: 4
## 5: 5
## 6: 6
## 7: 7
## 8: 8
## 9: 9
## 10: 10
# if we must specify the column names as characters
# set with to false in the argument
theDT[,"B", with=FALSE]
##
      В
## 1: a
## 2: b
## 3: c
## 4: d
## 5: e
## 6: f
## 7: g
## 8: h
## 9: i
## 10: j
theDT[, c("A","B"), with=FALSE]
##
       A B
## 1: 1 a
## 2: 2 b
## 3: 3 c
## 4: 4 d
## 5: 5 e
## 6: 6 f
## 7: 7 g
## 8: 8 h
## 9: 9 i
## 10: 10 j
# show tables
tables()
      NAME NROW NCOL MB COLS KEY
## 1: theDT
            10 4 0 A,B,C,D
## Total: OMB
class(theDT)
## [1] "data.table" "data.frame"
# The Key is used to index the data.table
# speed some calculation
```

```
# setup the key
setkey(theDT,D)
theDT
##
       A B C
## 1: 1 a K
             One
## 2: 4 d N One
## 3: 7 g Q
             One
## 4: 10 j T
             One
## 5: 3 c M Three
## 6: 6 f P Three
## 7: 9 i S Three
## 8: 2 b L
             Two
## 9: 5 e O
             Two
## 10: 8 h R
             Two
\# the data have been reordered according to column D
# sorted alphabatically
tables()
      NAME NROW NCOL MB COLS KEY
## 1: theDT
            10 4 0 A,B,C,D D
## Total: OMB
head(theDT)
     A B C
              D
## 1: 1 a K
## 2: 4 d N
              One
## 3: 7 g Q
              One
## 4: 10 j T
              One
## 5: 3 c M Three
## 6: 6 f P Three
# add more choices to select data
theDT["One",]
      A B C
##
## 1: 1 a K One
## 2: 4 d N One
## 3: 7 g Q One
## 4: 10 j T One
theDT[c("One","Three"),]
##
      A B C
## 1: 1 a K
              One
## 2: 4 d N
              One
## 3: 7 g Q
              One
## 4: 10 j T
              One
## 5: 3 c M Three
## 6: 6 f P Three
## 7: 9 i S Three
diamondsDT <- data.table(diamonds)</pre>
# set key for more than one column
```

```
setkey(diamondsDT, cut, color)
\# access some rows, the special function J
diamondsDT[J("Ideal", "E"),]
##
        carat
                cut color clarity depth table price
                                                      X
                                                            У
##
     1: 0.23 Ideal
                              SI2 61.5
                                                326 3.95 3.98 2.43
                        Ε
                                           55
##
     2: 0.26 Ideal
                        Ε
                             VVS2 62.9
                                           58
                                                554 4.02 4.06 2.54
                              SI1 62.5
##
     3: 0.70 Ideal
                                           57 2757 5.70 5.72 3.57
                        Ε
     4: 0.59 Ideal
                        Ε
                             VVS2 62.0
                                           55 2761 5.38 5.43 3.35
##
                              SI2 62.2
##
     5: 0.74 Ideal
                        Ε
                                           56 2761 5.80 5.84 3.62
##
## 3899: 0.70 Ideal
                                           55 2745 5.71 5.74 3.53
                        Ε
                              SI1 61.7
## 3900: 0.51 Ideal
                        Ε
                             VVS1
                                   61.9
                                           54 2745 5.17 5.11 3.18
## 3901: 0.56 Ideal
                        Ε
                             VVS1 62.1
                                           56 2750 5.28 5.29 3.28
## 3902: 0.77 Ideal
                        Ε
                              SI2 62.1
                                           56 2753 5.84 5.86 3.63
## 3903: 0.71 Ideal
                        Ε
                              SI1
                                   61.9
                                           56 2756 5.71 5.73 3.54
diamondsDT[J("Ideal",c("E", "D"))]
##
                cut color clarity depth table price
        carat
                                                      X
##
     1: 0.23 Ideal
                              SI2 61.5
                                                326 3.95 3.98 2.43
                        Ε
                                           55
##
     2: 0.26 Ideal
                        Ε
                             VVS2 62.9
                                           58
                                                554 4.02 4.06 2.54
     3: 0.70 Ideal
                              SI1 62.5
                                           57 2757 5.70 5.72 3.57
##
                        Ε
##
     4: 0.59 Ideal
                        Ε
                             VVS2 62.0
                                           55 2761 5.38 5.43 3.35
##
     5: 0.74 Ideal
                        Ε
                              SI2 62.2
                                           56 2761 5.80 5.84 3.62
##
## 6733: 0.51 Ideal
                        D
                             VVS2
                                   61.7
                                           56 2742 5.16 5.14 3.18
                             VVS2 61.3
## 6734: 0.51 Ideal
                                           57 2742 5.17 5.14 3.16
                        D
## 6735: 0.81 Ideal
                        D
                              SI1 61.5
                                           57 2748 6.00 6.03 3.70
## 6736: 0.72 Ideal
                        D
                              SI1 60.8
                                           57 2757 5.75 5.76 3.50
## 6737: 0.75 Ideal
                        D
                              SI2 62.2
                                           55 2757 5.83 5.87 3.64
# data.table aggregation
# the default aggregation function will be slower in data.table
# Let's have one comparison
aggregate(price ~ cut, diamonds, mean)
##
          cut
                 price
## 1
         Fair 4358.758
## 2
         Good 3928.864
## 3 Very Good 3981.760
     Premium 4584.258
## 4
        Ideal 3457.542
# in data.table
diamondsDT[, mean(price), by = cut] # notice that name becomes V1
##
           cut
## 1:
          Fair 4358.758
## 2:
          Good 3928.864
```

```
## 3: Very Good 3981.760
## 4:
        Premium 4584.258
## 5:
          Ideal 3457.542
# to fix the name issue
diamondsDT[,list(price = mean(price)), by = cut]
##
            \operatorname{cut}
                    price
## 1:
           Fair 4358.758
## 2:
           Good 3928.864
## 3: Very Good 3981.760
## 4:
        Premium 4584.258
## 5:
          Ideal 3457.542
# more column
diamondsDT[,list(price = mean(price)), by = list(cut, color)]
##
             cut color
                           price
##
   1:
            Fair
                      D 4291.061
## 2:
            Fair
                      E 3682.312
## 3:
                      F 3827.003
            Fair
## 4:
            Fair
                      G 4239.255
## 5:
            Fair
                      H 5135.683
## 6:
                      I 4685.446
            Fair
## 7:
            Fair
                      J 4975.655
## 8:
                      D 3405.382
            Good
## 9:
            Good
                      E 3423.644
## 10:
            Good
                      F 3495.750
## 11:
                      G 4123.482
            {\tt Good}
## 12:
            {\tt Good}
                      H 4276.255
## 13:
            Good
                      I 5078.533
## 14:
            Good
                      J 4574.173
## 15: Very Good
                      D 3470.467
## 16: Very Good
                      E 3214.652
## 17: Very Good
                      F 3778.820
## 18: Very Good
                      G 3872.754
## 19: Very Good
                      H 4535.390
## 20: Very Good
                      I 5255.880
## 21: Very Good
                      J 5103.513
## 22:
         Premium
                      D 3631.293
## 23:
         Premium
                      E 3538.914
## 24:
         Premium
                      F 4324.890
## 25:
         Premium
                      G 4500.742
## 26:
         Premium
                      H 5216.707
## 27:
         Premium
                      I 5946.181
## 28:
                      J 6294.592
         Premium
## 29:
           Ideal
                      D 2629.095
## 30:
                      E 2597.550
           Ideal
## 31:
           Ideal
                      F 3374.939
                      G 3720.706
## 32:
           Ideal
## 33:
           Ideal
                      H 3889.335
## 34:
           Ideal
                      I 4451.970
## 35:
           Ideal
                      J 4918.186
##
                           price
```

cut color

## # more aggregate diamondsDT[,list(price = mean(price), carat = mean(carat), caratsum = sum(carat)), by = list(cut, color ## cut color price carat caratsum ## Fair D 4291.061 0.9201227 1: 149.98 ## 2: Fair E 3682.312 0.8566071 191.88 ## 3: F 3827.003 0.9047115 282.27 Fair ## 4: G 4239.255 1.0238217 321.48 Fair ## 5: H 5135.683 1.2191749 369.41 Fair ## 6: I 4685.446 1.1980571 209.66 Fair ## 7: Fair J 4975.655 1.3411765 159.60 ## 8: Good D 3405.382 0.7445166 492.87 ## 9: E 3423.644 0.7451340 Good 695.21 ## 10: F 3495.750 0.7759296 ${\tt Good}$ 705.32 ## 11: G 4123.482 0.8508955 Good 741.13 ## 12: Good H 4276.255 0.9147293 642.14 ## 13: Good I 5078.533 1.0572222 551.87 ## 14: Good J 4574.173 1.0995440 337.56 ## 15: Very Good D 3470.467 0.6964243 1053.69 ## 16: Very Good E 3214.652 0.6763167 1623.16 ## 17: Very Good F 3778.820 0.7409612 1603.44 ## 18: Very Good G 3872.754 0.7667986 1762.87 ## 19: Very Good H 4535.390 0.9159485 1670.69 ## 20: Very Good I 5255.880 1.0469518 1260.53 ## 21: Very Good J 5103.513 1.1332153 768.32 ## 22: Premium D 3631.293 0.7215471 1156.64 ## 23: Premium E 3538.914 0.7177450 1677.37 ## 24: F 4324.890 0.8270356 1927.82 Premium ## 25: G 4500.742 0.8414877 Premium 2460.51 ## 26: H 5216.707 1.0164492 2398.82 Premium ## 27: Premium I 5946.181 1.1449370 1634.97 ## 28: Premium J 6294.592 1.2930941 1044.82 ## 29: Ideal D 2629.095 0.5657657 1603.38 ## 30: Ideal E 2597.550 0.5784012 2257.50 ## 31: Ideal F 3374.939 0.6558285 2509.20 ## 32: G 3720.706 0.7007146 3422.29 Ideal

## 33:

## 34:

## 35:

##

Ideal

Ideal

Ideal

cut color

Н 3889.335 0.7995249

I 4451.970 0.9130291

J 4918.186 1.0635937

price

2490.52

1910.97

carat caratsum

952.98