Fun with functions and dplyr

Brian Wright

1/24/2020

Overview of Functions (Advanced R)

 Functions are at the core of R language, it's really a function based language

"R, at its heart, is a functional language. This means that it has certain technical properties, but more importantly that it lends itself to a style of problem solving centred on functions." Hadley Wickham

What is a functional based language?

- Recently functions have grown in popularity because they can produce efficient and simple solutions to lots of problems. Many of the problems with performance have been solved.
- Functional programming compliments object oriented programming

What makes a programming approach "functional"?

- Functions can behave like any other data structure
 - Assign them to variables, store to lists, pass them as aurguments to other functions, create them inside functions and even produce a function as a result of a funcion
- Functions need to be "pure" meaning that if you call it again with the same inputs you get the same results. sys.time() not a "pure" function
- The execution of the function shouldn't change global variables, have no side effects.

Functions

- Function don't have to be "pure" but it can help to ensure your code is doing what you intend it to do.
- Functional programming helps to break a problem down into it's pieces. When working to solve a problem it helps to divide the code into individually operating functions that solve parts of the problem.

Types of Functions

In Out	Vector	Function
Vector	Regular function	Function factory
Function	Functional	Function operator

Figure 1: Function Types

Let's Build a Function

• Basically recipes composed of series of R statements

```
name <- funtion(variables){
    #In here goes the series of R statements
}</pre>
```

Example, talk out the steps

```
my_mean <- function(x){
   Sum <- sum(x) #Here we are using a function
   #inside a function!
   N <- length(x)
   return(Sum/N) #return is optional but helps with
   #clarity on some level.
}</pre>
```

Create a little list and pass it to the function and see if it works. Also call the Sum and N variables. . . does this work?

Functional - Will show later, Function Factory (Advanced R)

```
power1 <- function(exp) {
  function(x) {
    x ^ exp
  }
}

#Assigning the exponentials
square <- power1(2)
cube <- power1(3)</pre>
```

Run the Created Functions

square(3)

> [1] 9

cube(3)

> [1] 27

Quick Exercise

Create a function that computes the range of a variable and then for no good reason adds 100 and divides by 10. Write out the steps you would need first in Pseudocode, then develop the function.

dplyr verbs in the tidyverse

The dplyr package gives us a few verbs for data manipulation

Function	Purpose
select	Select columns based on name or position
mutate	Create or change a column
filter	Extract rows based on some criteria
arrange	Re-order rows based on values of variable(s)
group_by	Split a dataset by unique values of a variable
summarize	Create summary statistics based on columns

select

You can select columns by name or position, of course.

You can also select columns based on some criteria, which are encapsulated in functions.

- starts_with(""), ends_with(""), contains("____")
- one_of("____","____","____")

There are others; see help(starts_with).

Example

Load the weather.csv. This contains daily temperature data in 2010 for some location.

```
head(weather, 2)
```

A tibble: 2 x 35

```
>
   id year month element d1 d2 d3 d4
                                              d5
>
   <chr> <int> <int> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <
>
  1 MX17~ 2010 1 tmax
                           NA NA
                                    NA NA
                                              NA
>
  # ... with 23 more variables: d9 <lgl>, d10 <dbl>, d11 <dbl
     d13 <dbl>, d14 <dbl>, d15 <dbl>, d16 <dbl>, d17 <dbl>,
>
     d19 <lgl>, d20 <lgl>, d21 <lgl>, d22 <lgl>, d23 <dbl>,
>
    d25 <dbl>, d26 <dbl>, d27 <dbl>, d28 <dbl>, d29 <dbl>,
     d31 <dbl>
```

How would you just select the columns with the daily data?

```
select(weather, starts with("d"))
```

mutate

mutate can either transform a column in place or create a new column in a dataset

We'll use the in-built mpg dataset for this example, We'll select only the city and highway mileages. To use this selection later, we will need to assign it to a new name

```
mpg1 <- select(mpg, cty, hwy)</pre>
```

mutate

We'll change the city and highway mileage to km/l from mpg. This will involve multiplying it by 1.6 and dividing by 3.8

```
head(mutate(mpg1, cty = cty * 1.6 / 3.8,
hwy = hwy * 1.6/3.8), 5)
```

```
> # A tibble: 5 x 2
> cty hwy
> <dbl> <dbl>>
> 1 7.58 12.2
> 2 8.84 12.2
> 3 8.42 13.1
> 4 8.84 12.6
> 5 6.74 10.9
```

This is in-place replacement

New Variable Defined

```
mutate(mpg1, cty1 = cty * 1.6/3.8, hwy1 = hwy * 1.6/3.8)
```

```
>
     A tibble: 234 \times 4
>
        cty
               hwy cty1
                           hwy1
>
      <int> <int> <dbl> <dbl>
         18
                29
                    7.58
                           12.2
>
>
         21
                29
                    8.84
                           12.2
    3
>
         20
                31 8.42
                           13.1
>
         21
                30 8.84
                           12.6
    5
>
         16
                26
                   6.74
                           10.9
>
    6
         18
                26 7.58
                           10.9
>
    7
         18
                27
                   7.58
                           11.4
>
    8
         18
                    7.58
                26
                           10.9
    9
         16
>
                25
                    6.74
                           10.5
>
   10
         20
                28
                    8.42
                           11.8
>
     ... with 224 more rows
```

This creates new variables

filter

filter extracts rows based on criteria

```
filter(mpg, cyl == 4)
   # A tibble: 81 x 11
>
     manufacturer model displ year cyl trans drv
>
                                                         cty
                   <chr> <dbl> <int> <int> <chr> <chr> <int> <
>
     <chr>
    1 audi
                           1.8
                                                          18
>
                   a4
                                1999
                                         4 auto~ f
>
   2 audi
                   a4
                           1.8
                                1999
                                         4 manu~ f
                                                          21
   3 audi
                           2
                                2008
                                                          20
>
                   a4
                                         4 manu~ f
                          2
                                                          21
>
   4 audi
                   a4
                                2008
                                         4 auto~ f
                                                          18
>
   5 audi
                  a4 q~ 1.8 1999
                                         4 manu~ 4
                                                          16
>
   6 audi
                  a4 q~ 1.8
                               1999
                                         4 auto~ 4
>
   7 audi
                   a4 q~
                          2
                                2008
                                         4 manu~ 4
                                                          20
                          2 2008
>
   8 audi
                   a4 q~
                                         4 auto~ 4
                                                          19
>
    9 chevrolet
                   mali~
                           2.4 1999
                                         4 auto~ f
                                                          19
>
   10 chevrolet
                   mali~
                           2.4
                                2008
                                         4 auto~ f
                                                          22
        with 71 more rows
```

Brian Wright

Practice Piping

```
admit_df <- read_csv("LogReg.csv")
str(admit df)
   Classes 'spec_tbl_df', 'tbl_df', 'tbl' and 'data.frame': 40
>
   $ admit: num 0 1 1 1 0 1 1 0 1 0 ...
   $ gre : num 380 660 800 640 520 760 560 400 540 700 ...
>
   $ gpa : num 3.61 3.67 4 3.19 2.93 3 2.98 3.08 3.39 3.92
>
>
   $ rank : num 3 3 1 4 4 2 1 2 3 2 ...
>
    - attr(*, "spec")=
>
     .. cols(
>
       admit = col double(),
>
     .. gre = col double(),
>
     .. gpa = col double(),
>
     .. rank = col double()
>
```

#Do we notice anything that seems a bit off.

Coercion num to factor

```
admit_df$rank <- as.factor(admit_df$rank)
#changes rank to a factor</pre>
```

Five Basic Classes in R

- character
- numeric (double precision floating point numbers, default)
- integer (subset of numeric)
- complex (j = 10 + 5i)
- logical (True/False)

All have coercion calls (example from: R Nuts and Bolts)

```
x <- 0:6
class(x)#why
```

> [1] "integer"

```
as.numeric(x)
```

> [1] 0 1 2 3 4 5 6

```
as.logical(x)
```

> [1] FALSE TRUE TRUE TRUE TRUE TRUE TRUE

```
as.character(x)
```

> [1] "0" "1" "2" "3" "4" "5" "6"

Functional Example: Pass a function get a vector

We can also convert multiple columns using lapply(), great example of functional orientation of R.

> [1] "factor" "factor"

#using a functional with two functions inside that creates a object coerced to a character list... what fun.

Using the code chunk below to "group_by" rank

Using the code chunk below to filter by 1 in the admit column

Ok now summarise by average GPA

Now Pipe everything together