



Data Science for Environmental Health



Functions

Writing your own functions

So far we've seen many functions, like `c()`, `class()`, `filter()`, `dim()` ...

Why create your own functions?

- Cut down on repetitive code (easier to fix things!)
- Organize code into manageable chunks
- Avoid running code unintentionally
- Use names that make sense to you

Writing your own functions

The general syntax for a function is:

```
function_name <- function(arg1, arg2, ...) {  
  <function body>  
}
```

Writing your own functions

Here we will write a function that multiplies some number x by 2:

```
div_100 <- function(x) x / 100
```

When you run the line of code above, you make it ready to use (no output yet!).
Let's test it!

```
div_100(x = 600)
```

```
[1] 6
```

Writing your own functions: { }

Adding the curly brackets - {} - allows you to use functions spanning multiple lines:

```
div_100 <- function(x) {  
  x / 100  
}  
div_100(x = 10)
```

```
[1] 0.1
```

Writing your own functions: **return**

If we want something specific for the function's output, we use **return()**:

```
div_100_plus_4 <- function(x) {  
  output_int <- x / 100  
  output <- output_int + 4  
  return(output)  
}  
div_100_plus_4(x = 10)
```

```
[1] 4.1
```

Writing your own functions: multiple inputs

Functions can take multiple inputs:

```
div_100_plus_y <- function(x, y) x / 100 + y  
div_100_plus_y(x = 10, y = 3)
```

```
[1] 3.1
```

Writing your own functions: multiple outputs

Functions can return a vector (or other object) with multiple outputs.

```
x_and_y_plus_2 <- function(x, y) {  
  output1 <- x + 2  
  output2 <- y + 2  
  
  return(c(output1, output2))  
}  
result <- x_and_y_plus_2(x = 10, y = 3)  
result
```

```
[1] 12  5
```

Writing your own functions: defaults

Functions can have “default” arguments. This lets us use the function without using an argument later:

```
div_100_plus_y <- function(x = 10, y = 3) x / 100 + y  
div_100_plus_y()
```

```
[1] 3.1
```

```
div_100_plus_y(x = 11, y = 4)
```

```
[1] 4.11
```

Writing another simple function

Let's write a function, `sqdif`, that:

1. takes two numbers `x` and `y` with default values of 2 and 3.
2. takes the difference
3. squares this difference
4. then returns the final value

Writing another simple function

```
sqdif <- function(x = 2, y = 3) (x - y)^2
```

```
sqdif()
```

```
[1] 1
```

```
sqdif(x = 10, y = 5)
```

```
[1] 25
```

```
sqdif(10, 5)
```

```
[1] 25
```

```
sqdif(11, 4)
```

```
[1] 49
```

Writing your own functions: characters

Functions can have any kind of input. Here is a function with characters:

```
loud <- function(word) {  
  output <- rep(toupper(word), 5)  
  return(output)  
}  
loud(word = "hooray!")  
  
[1] "HOORAY!" "HOORAY!" "HOORAY!" "HOORAY!" "HOORAY!"
```

Functions for tibbles - curly braces

```
# get means and missing for a specific column
get_summary <- function(dataset, col_name) {
  dataset %>%
    summarise(mean = mean({{col_name}}), na.rm = TRUE),
    na_count = sum(is.na({{col_name}})))
}
```

Examples:

```
get_summary(calenviroscreen, CES4.0Score)
```

```
# A tibble: 1 × 2
  mean na_count
  <dbl>   <int>
1  28.3     103
```

```
get_summary(haa5, perc_pop_exposed_to_exceedances)
```

```
# A tibble: 1 × 2
  mean na_count
  <dbl>   <int>
1 0.0591      11
```

Summary

- Simple functions take the form:
 - `NEW_FUNCTION <- function(x, y){x + y}`
 - Can specify defaults like `function(x = 1, y = 2){x + y}` -`return` will provide a value as output
 - `print` will simply print the value on the screen but not save it

Lab Part 1

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Functions on multiple columns

Using your custom functions: **sapply()**- a base R function

Now that you've made a function... You can "apply" functions easily with **sapply()**!

These functions take the form:

```
sapply(<a vector, list, data frame>, some_function)
```

CalEnviroScreen

This dataset was gathered by the California Office of Environmental Health Hazard Assessment. CalEnviroScreen ranks census tracts in California based on potential exposures to pollutants, adverse environmental conditions, socioeconomic factors and the prevalence of certain health conditions. Read more at <https://calenviroscreen-oehha.hub.arcgis.com/>.

```
head(calenviroscreen)
```

```
# A tibble: 6 × 67
  CensusTract CaliforniaCounty    ZIP Longitude Latitude ApproxLocation
  <dbl> <chr>          <int>     <dbl>      <dbl> <chr>
1 6001400100 "Alameda"        94704     -122.      37.9 Oakland
2 6001400200 "Alameda"        94618     -122.      37.8 Oakland
3 6001400300 "Alameda"        94618     -122.      37.8 Oakland
4 6001400400 "Alameda"        94609     -122.      37.8 Oakland
5 6001400500 "Alameda"        94609     -122.      37.8 Oakland
6 6001400600 "Alameda"        94609     -122.      37.8 Oakland
# i 61 more variables: CES4.0Score <dbl>, CES4.0Percentile <dbl>,
# CES4.0PercRange <chr>, Ozone <dbl>, OzonePctl <dbl>, PM2.5 <dbl>,
# PM2.5.Pctl <dbl>, DieselPM <dbl>, DieselPMPctl <dbl>, DrinkingWater <dbl>,
# DrinkingWaterPctl <dbl>, Lead <dbl>, LeadPctl <dbl>, Pesticides <dbl>,
# PesticidesPctl <dbl>, ToxRelease <dbl>, ToxReleasePctl <dbl>,
# Traffic <dbl>, TrafficPctl <dbl>, CleanupSites <dbl>,
# CleanupSitesPctl <dbl>, GroundwaterThreats <dbl>, ...
```

Using your custom functions: `sapply()`

There are no parentheses on the functions!

You can also pipe into your function.

```
sapply(calenviroscreen, class) # also: calenviroscreen %>% sapply(class)
```

CensusTract	CaliforniaCounty	ZIP
"numeric"	"character"	"integer"
Longitude	Latitude	ApproxLocation
"numeric"	"numeric"	"character"
CES4.0Score	CES4.0Percentile	CES4.0PercRange
"numeric"	"numeric"	"character"
Ozone	OzonePctl	PM2.5
"numeric"	"numeric"	"numeric"
PM2.5.Pctl	DieselPM	DieselPMPctl
"numeric"	"numeric"	"numeric"
DrinkingWater	DrinkingWaterPctl	Lead
"numeric"	"numeric"	"numeric"
LeadPctl	Pesticides	PesticidesPctl
"numeric"	"numeric"	"numeric"
ToxRelease	ToxReleasePctl	Traffic
"numeric"	"numeric"	"numeric"
TrafficPctl	CleanupSites	CleanupSitesPctl
"numeric"	"numeric"	"numeric"
GroundwaterThreats	GroundwaterThreatsPctl	HazWaste
"numeric"	"numeric"	"numeric"
HazWastePctl	ImpWaterBodies	ImpWaterBodiesPctl
"numeric"	"integer"	"numeric"
SolidWaste	SolidWastePctl	PollutionBurden
"numeric"	"numeric"	"numeric"

Using your custom functions: `sapply()`

Use the `div_100` function we created earlier to convert 0-100 percentiles to proportions.

```
calenviroscreen %>%
  select(ends_with("Pct1")) %>%
  sapply(div_100) %>%
  head()
```

	OzonePctl	PM2.5.Pctl	DieselPMPctl	DrinkingWaterPctl	LeadPctl
[1,]	0.0312	0.3627	0.3476	0.0421	0.0774
[2,]	0.0312	0.4197	0.9271	0.0421	0.6820
[3,]	0.0312	0.4390	0.8977	0.0421	0.6418
[4,]	0.0312	0.4281	0.7910	0.0421	0.6708
[5,]	0.0312	0.4281	0.6758	0.0421	0.6795
[6,]	0.0312	0.4281	0.8376	0.0421	0.6970
	PesticidesPctl	ToxReleasePctl	TrafficPctl	CleanupSitesPctl	
[1,]	0	0.5603	0.5594	0.5817	
[2,]	0	0.5543	0.3749	0.0000	
[3,]	0	0.5504	0.4248	0.1183	
[4,]	0	0.5590	0.3800	0.0000	
[5,]	0	0.5648	0.4868	0.3387	
[6,]	0	0.5565	0.6706	0.2262	
	GroundwaterThreatsPctl	HazWastePctl	ImpWaterBodiesPctl	SolidWastePctl	
[1,]	0.5242	0.9252	0.2388	0.3572	
[2,]	0.8793	0.2851	0.0000	0.0000	
[3,]	0.8529	0.7407	0.0000	0.0000	
[4,]	0.9256	0.5189	0.0000	0.0000	
[5,]	0.8434	0.5640	0.0000	0.0000	
[6,]	0.7906	0.5827	0.0000	0.0000	

Using your custom functions “on the fly” to iterate

Also called “anonymous function”.

```
calenviroscreen %>%
  select(ends_with("Pct1")) %>%
  sapply(function(x) x / 100) %>%
  head()
```

	OzonePct1	PM2.5.Pct1	DieselPMPct1	DrinkingWaterPct1	LeadPct1
[1,]	0.0312	0.3627	0.3476	0.0421	0.0774
[2,]	0.0312	0.4197	0.9271	0.0421	0.6820
[3,]	0.0312	0.4390	0.8977	0.0421	0.6418
[4,]	0.0312	0.4281	0.7910	0.0421	0.6708
[5,]	0.0312	0.4281	0.6758	0.0421	0.6795
[6,]	0.0312	0.4281	0.8376	0.0421	0.6970
	PesticidesPct1	ToxReleasePct1	TrafficPct1	CleanupSitesPct1	
[1,]	0	0.5603	0.5594	0.5817	
[2,]	0	0.5543	0.3749	0.0000	
[3,]	0	0.5504	0.4248	0.1183	
[4,]	0	0.5590	0.3800	0.0000	
[5,]	0	0.5648	0.4868	0.3387	
[6,]	0	0.5565	0.6706	0.2262	
	GroundwaterThreatsPct1	HazWastePct1	ImpWaterBodiesPct1	SolidWastePct1	
[1,]	0.5242	0.9252	0.2388	0.3572	
[2,]	0.8793	0.2851	0.0000	0.0000	
[3,]	0.8529	0.7407	0.0000	0.0000	
[4,]	0.9256	0.5189	0.0000	0.0000	
[5,]	0.8434	0.5640	0.0000	0.0000	
[6,]	0.7906	0.5827	0.0000	0.0000	
	PollutionBurdenPct1	AsthmaPct1	LowBirthWeightPct1		/
[1,]	0.2662	0.0444	0.2306		

Anonymous functions: alternative syntax

```
calenviroscreen %>%
  select(ends_with("Pct1")) %>%
  sapply(\(x) x / 100) %>%
  head()
```

	OzonePct1	PM2.5.Pct1	DieselPMPct1	DrinkingWaterPct1	LeadPct1
[1,]	0.0312	0.3627	0.3476	0.0421	0.0774
[2,]	0.0312	0.4197	0.9271	0.0421	0.6820
[3,]	0.0312	0.4390	0.8977	0.0421	0.6418
[4,]	0.0312	0.4281	0.7910	0.0421	0.6708
[5,]	0.0312	0.4281	0.6758	0.0421	0.6795
[6,]	0.0312	0.4281	0.8376	0.0421	0.6970
	PesticidesPct1	ToxReleasePct1	TrafficPct1	CleanupSitesPct1	
[1,]	0	0.5603	0.5594	0.5817	
[2,]	0	0.5543	0.3749	0.0000	
[3,]	0	0.5504	0.4248	0.1183	
[4,]	0	0.5590	0.3800	0.0000	
[5,]	0	0.5648	0.4868	0.3387	
[6,]	0	0.5565	0.6706	0.2262	
	GroundwaterThreatsPct1	HazWastePct1	ImpWaterBodiesPct1	SolidWastePct1	
[1,]	0.5242	0.9252	0.2388	0.3572	
[2,]	0.8793	0.2851	0.0000	0.0000	
[3,]	0.8529	0.7407	0.0000	0.0000	
[4,]	0.9256	0.5189	0.0000	0.0000	
[5,]	0.8434	0.5640	0.0000	0.0000	
[6,]	0.7906	0.5827	0.0000	0.0000	
	PollutionBurdenPct1	AsthmaPct1	LowBirthWeightPct1		/
[1,]	0.2662	0.0444	0.2306		
[2,]	0.2418	0.0980	0.2792		
[3,]	0.3337	0.2657	0.2162		
[4,]	0.2684	0.5568	0.2788		

across

Using functions in `mutate()` and `summarize()`

Already know how to use functions to modify columns using `mutate()` or calculate summary statistics using `summarize()`.

- Pesticides - pounds of selected active pesticide / square mile
- Poverty - percent of population living below two times the federal poverty level
- LowBirthWeight - Percent low birth weight

```
calenviroscreen %>%
  summarize(max_pest = max(Pesticides, na.rm = T),
           max_pov = max(Poverty, na.rm = T),
           low_bw = max(LowBirthWeight, na.rm = T))
```

```
# A tibble: 1 × 3
  max_pest max_pov low_bw
  <dbl>    <dbl>   <dbl>
1     80811.    96.7    13.7
```

Applying functions with **across** from **dplyr**

`across()` makes it easy to apply the same transformation to multiple columns.
Usually used with `summarize()` or `mutate()`.

```
summarize(across(<columns>, function))
```

or

```
mutate(across(<columns>, function))
```

- List columns first: `.cols =`
- List function next: `.fns =`
- If there are arguments to a function (e.g., `na.rm = TRUE`), use an anonymous function.

Applying functions with `across` from `dplyr`

Combining with `summarize()`

```
calenviroscreen %>%  
  summarize(across(  
    c(Pesticides, Poverty, LowBirthWeight),  
    mean # no parentheses  
  ))
```

```
# A tibble: 1 × 3  
  Pesticides Poverty LowBirthWeight  
        <dbl>     <dbl>          <dbl>  
1       268.      NA            NA
```

Applying functions with **across** from **dplyr**

Add anonymous function to include additional arguments (e.g., `na.rm = T`).

```
calenviroscreen %>%  
  summarize(across(  
    c(Pesticides, Poverty, LowBirthWeight),  
    function(x) mean(x, na.rm = T)  
)
```

```
# A tibble: 1 × 3  
  Pesticides Poverty LowBirthWeight  
        <dbl>     <dbl>          <dbl>  
1       268.     31.3          5.00
```

Applying functions with `across` from `dplyr`

Can use with other tidyverse functions like `group_by`!

```
calenviroscreen %>%
  group_by(CaliforniaCounty) %>%
  summarize(across(
    c(Pesticides, Poverty, LowBirthWeight),
    function(x) mean(x, na.rm = T)
  ))
```

```
# A tibble: 58 × 4
  CaliforniaCounty Pesticides Poverty LowBirthWeight
  <chr>           <dbl>     <dbl>      <dbl>
1 "Alameda "       0.948     22.1       5.26
2 "Alpine "        0          38.9       NaN
3 "Amador "        2.01      25.2       4.45
4 "Butte "         736.      39.9       4.64
5 "Calaveras "     1.20      28.2       3.55
6 "Colusa "        1186.     36.5       4.11
7 "Contra Costa"   10.5      20.6       4.71
8 "Del Norte"      47.4      48.4       NaN
9 "El Dorado"      5.50      20.9       4.28
10 "Fresno "       586.      45.8       5.96
# i 48 more rows
```

Applying functions with `across` from `dplyr`

Using different `tidyselect()` options (e.g., `starts_with()`, `ends_with()`, `contains()`)

```
calenviroscreen %>%  
  group_by(CaliforniaCounty) %>%  
  summarize(across(contains("PM"), mean))
```

```
# A tibble: 58 × 5  
  CaliforniaCounty PM2.5 PM2.5.Pctl DieselPM DieselPMPctl  
  <chr>          <dbl>    <dbl>     <dbl>      <dbl>  
1 "Alameda "      8.87     31.9     0.350     66.4  
2 "Alpine "       3.05     0.07     0.003     1.02  
3 "Amador "       8.01     18.9     0.0111    4.18  
4 "Butte "        8.22     23.0     0.106     33.1  
5 "Calaveras "    8.12     22.6     0.0079    3.08  
6 "Colusa "       7.54     13.3     0.0292    10.6  
7 "Contra Costa" 8.76     31.0     0.210     48.6  
8 "Del Norte"    5.71     2.93     0.0301    11.1  
9 "El Dorado"    6.78     8.91     0.0380    13.6  
10 "Fresno "      13.2     91.4     0.181     44.8  
# i 48 more rows
```

Applying functions with `across` from `dplyr`

Combining with `mutate()` - the `replace_na` function

Here we will use the `yearly_co2_emissions` data from `dasehr`

`replace_na({data frame}, {list of values})` or `replace_na({vector}, {single value})`

```
yearly_co2_emissions %>%
  select(country, starts_with("194")) %>%
  mutate(across(
    c(`1943`, `1944`, `1945`),
    function(x) replace_na(x, replace = 0)
  ))
```



```
# A tibble: 192 × 11
  country      `1940` `1941` `1942` `1943` `1944` `1945` `1946` `1947` `1948`
  <chr>       <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
1 Afghanistan     NA     NA     NA     0     0     0     NA     NA     NA
2 Albania        693    627    744    462    154    121    484    928    704
3 Algeria         238    312    499    469    499    616    763    744    803
4 Andorra         NA     NA     NA     0     0     0     NA     NA     NA
5 Angola          NA     NA     NA     0     0     0     NA     NA     NA
6 Antigua and B... NA     NA     NA     0     0     0     NA     NA     NA
7 Argentina       15900  14000  13500  14100  14000  13700  13700  14500  17400
8 Armenia          848    745    513    655    613    649    730    878    935
9 Australia       29100  34600  36500  35000  34200  32700  35500  38000  38500
10 Austria         7350   7980   8560   9620   9400   4570   12800  17600  24500
# i 182 more rows
# i 1 more variable: `1949` <dbl>
```

purrr package

Similar to `across`, `purrr` is a package that allows you to apply a function to multiple columns in a data frame or multiple data objects in a list.

While we won't get into `purrr` too much in this class, its a handy package for you to know about should you get into a situation where you have an irregular list you need to handle!

Multiple Data Frames

Multiple data frames

Lists help us work with multiple data frames

```
AQ_list <- list(AQ1 = airquality, AQ2 = airquality, AQ3 = airquality)
str(AQ_list)
```

```
List of 3
$ AQ1:'data.frame':    153 obs. of   6 variables:
..$ Ozone   : int [1:153] 41 36 12 18 NA 28 23 19 8 NA ...
..$ Solar.R: int [1:153] 190 118 149 313 NA NA 299 99 19 194 ...
..$ Wind    : num [1:153] 7.4 8 12.6 11.5 14.3 14.9 8.6 13.8 20.1 8.6 ...
..$ Temp    : int [1:153] 67 72 74 62 56 66 65 59 61 69 ...
..$ Month   : int [1:153] 5 5 5 5 5 5 5 5 5 ...
..$ Day     : int [1:153] 1 2 3 4 5 6 7 8 9 10 ...
$ AQ2:'data.frame':    153 obs. of   6 variables:
..$ Ozone   : int [1:153] 41 36 12 18 NA 28 23 19 8 NA ...
..$ Solar.R: int [1:153] 190 118 149 313 NA NA 299 99 19 194 ...
..$ Wind    : num [1:153] 7.4 8 12.6 11.5 14.3 14.9 8.6 13.8 20.1 8.6 ...
..$ Temp    : int [1:153] 67 72 74 62 56 66 65 59 61 69 ...
..$ Month   : int [1:153] 5 5 5 5 5 5 5 5 5 ...
..$ Day     : int [1:153] 1 2 3 4 5 6 7 8 9 10 ...
$ AQ3:'data.frame':    153 obs. of   6 variables:
..$ Ozone   : int [1:153] 41 36 12 18 NA 28 23 19 8 NA ...
..$ Solar.R: int [1:153] 190 118 149 313 NA NA 299 99 19 194 ...
..$ Wind    : num [1:153] 7.4 8 12.6 11.5 14.3 14.9 8.6 13.8 20.1 8.6 ...
..$ Temp    : int [1:153] 67 72 74 62 56 66 65 59 61 69 ...
..$ Month   : int [1:153] 5 5 5 5 5 5 5 5 5 ...
..$ Day     : int [1:153] 1 2 3 4 5 6 7 8 9 10 ...
```

Multiple data frames: `sapply`

```
AQ_list %>% sapply(class)
```

```
      AQ1          AQ2          AQ3  
"data.frame" "data.frame" "data.frame"
```

```
AQ_list %>% sapply(nrow)
```

```
AQ1 AQ2 AQ3  
153 153 153
```

```
AQ_list %>% sapply(colMeans, na.rm = TRUE)
```

	AQ1	AQ2	AQ3
Ozone	42.129310	42.129310	42.129310
Solar.R	185.931507	185.931507	185.931507
Wind	9.957516	9.957516	9.957516
Temp	77.882353	77.882353	77.882353
Month	6.993464	6.993464	6.993464
Day	15.803922	15.803922	15.803922

Summary

- Apply your functions with `sapply(<a vector or list>, some_function)`
- Use `across()` to apply functions across multiple columns of data
- Need to use `across` within `summarize()` or `mutate()`
- Can use `sapply` or `purrr` to work with multiple data frames within lists simultaneously

Lab Part 2

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