



R Functions

# Why functional programming?

# Vanilla cupcakes

## Ingredients:

1. Flour
2. Sugar
3. Baking powder
4. Unsalted butter
5. Milk
6. Egg
7. Vanilla

## Directions:

1. Preheat oven to 350°F
2. Put the flour, sugar, baking powder, salt, and butter in a free standing electric mixer with a paddle attachment, beat on slow speed until sandy consistency is obtained
3. Whisk ingredients 5-7 together
4. Spoon batter, bake for 20 minutes

# Chocolate cupcakes

## Ingredients:

1. Cocoa
2. Sugar
3. Baking powder
4. Unsalted butter
5. Milk
6. Egg
7. Vanilla

## Directions:

1. Preheat oven to 350°F
2. Put the cocoa, sugar, baking powder, salt, and butter in a free standing electric mixer with a paddle attachment, beat on slow speed until sandy consistency is obtained Whisk ingredients 6-8 together
3. Spoon batter, bake for 20 minutes

# Chocolate cupcakes

## Ingredients:

1. Cocoa
2. Sugar
3. Baking powder
4. Unsalted butter
5. Milk
6. Egg
7. Vanilla

## Directions:

1. Preheat oven to 350°F
2. Put the cocoa, sugar, baking powder, salt, and butter in a free standing electric mixer with a paddle attachment, beat on slow speed until sandy consistency is obtained Whisk ingredients 6-8 together
3. Spoon batter, bake for 20 minutes

# Vanilla cupcakes

## 1. Rely on domain knowledge

### Ingredients:

1. Flour
2. Sugar
3. Baking powder
4. Unsalted butter
5. Milk
6. Egg
7. Vanilla

### Directions:

1. Preheat oven to 350°F
2. Put the flour, sugar, baking powder, salt, and butter in a free standing electric mixer with a paddle attachment, beat on slow speed until sandy consistency is obtained
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4. Spoon batter, bake for 20 minutes

# Vanilla cupcakes

1. Rely on domain knowledge

## Ingredients:

1. Flour
2. Sugar
3. Baking powder
4. Unsalted butter
5. Milk
6. Egg
7. Vanilla

## Directions:

1. Preheat
2. Mix, whisk, and spoon
3. Bake

# Vanilla cupcakes

## 2. Use variables

### Ingredients:

1. Flour
2. Sugar
3. Baking powder
4. Unsalted butter
5. Milk
6. Egg
7. Vanilla

### Directions:

1. Preheat
2. Mix, whisk, and spoon
3. Bake

# Vanilla cupcakes

## 2. Use variables

### Ingredients:

1. Flour
2. Sugar
3. Baking powder
4. Unsalted butter
5. Milk
6. Egg
7. Vanilla

### Directions:

1. Preheat
2. Mix dry ingredients, whisk wet ingredients, and spoon
3. Bake



# Cupcakes

## Directions:

1. Preheat
2. Mix **dry ingredients**, whisk **wet ingredients**, and spoon
3. Bake

## 3. Extract out common code

### Vanilla:

1. **Flour**
2. **Sugar**
3. **Baking powder**
4. **Unsalted butter**
5. **Milk**
6. **Egg**
7. **Vanilla**

### Chocolate:

1. **Cocoa**
2. **Sugar**
3. **Baking powder**
4. **Unsalted butter**
5. **Milk**
6. **Egg**
7. **Vanilla**



# for loops are like pages in the recipe book

```
> out1 <- vector("double", ncol(mtcars))

for(i in seq_along(mtcars)) {
  out1[[i]] <- mean(mtcars[[i]], na.rm = TRUE)
}

> out2 <- vector("double", ncol(mtcars))

for(i in seq_along(mtcars)) {
  out2[[i]] <- median(mtcars[[i]], na.rm = TRUE)
}
```

# for loops are like pages in the recipe book

```
> out1 <- vector("double", ncol(mtcars))

for(i in seq_along(mtcars)) {
  out1[[i]] <- mean(mtcars[[i]], na.rm = TRUE)
}

> out2 <- vector("double", ncol(mtcars))

for(i in seq_along(mtcars)) {
  out2[[i]] <- median(mtcars[[i]], na.rm = TRUE)
}
```

- Emphasizes the objects, pattern of implementation
- Hides actions

# for loops are like pages in the recipe book

```
> out1 <- vector("double", ncol(mtcars))

for(i in seq_along(mtcars)) {
  out1[[i]] <- mean(mtcars[[i]], na.rm = TRUE)
}

> out2 <- vector("double", ncol(mtcars))

for(i in seq_along(mtcars)) {
  out2[[i]] <- median(mtcars[[i]], na.rm = TRUE)
}
```

- Emphasizes the objects, pattern of implementation
- Hides actions

# Functional programming is like the meta-recipe

```
> library(purrr)

> means <- map_dbl(mtcars, mean)

> medians <- map_dbl(mtcars, median)
```

- Give equal weight to verbs and nouns
- Abstract away the details of implementation



R Functions

**Let's practice!**



R Functions

**Functions can be  
arguments too**

# Removing duplication with arguments

```
> f1 <- function(x) abs(x - mean(x)) ^ 1  
> f2 <- function(x) abs(x - mean(x)) ^ 2  
> f3 <- function(x) abs(x - mean(x)) ^ 3
```



# Removing duplication with arguments

```
> f1 <- function(x) abs(x - mean(x)) ^ power
> f2 <- function(x) abs(x - mean(x)) ^ power
> f3 <- function(x) abs(x - mean(x)) ^ power
```

# Removing duplication with arguments

```
> f1 <- function(x, power) abs(x - mean(x)) ^ power  
> f2 <- function(x, power) abs(x - mean(x)) ^ power  
> f3 <- function(x, power) abs(x - mean(x)) ^ power
```

# Functions can be arguments too

```
col_median <- function(df) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- median(df[[i]])  
  }  
  output  
}
```

```
col_sd <- function(df) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- sd(df[[i]])  
  }  
  output  
}
```

```
col_mean <- function(df) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- mean(df[[i]])  
  }  
  output  
}
```

# Functions can be arguments too

```
col_median <- function(df) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- fun(df[[i]])  
  }  
  output  
}
```

```
col_sd <- function(df) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- fun(df[[i]])  
  }  
  output  
}
```

```
col_mean <- function(df) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- fun(df[[i]])  
  }  
  output  
}
```

# Functions can be arguments too

```
col_median <- function(df, fun) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- fun(df[[i]])  
  }  
  output  
}
```

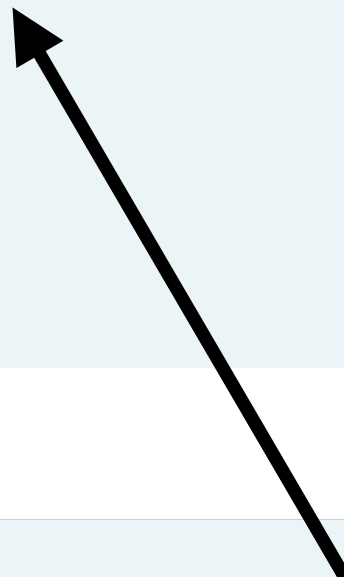
```
col_sd <- function(df, fun) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- fun(df[[i]])  
  }  
  output  
}
```

```
col_mean <- function(df, fun) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- fun(df[[i]])  
  }  
  output  
}
```

# Functions can be arguments too

```
col_summary <- function(df, fun) {  
  output <- numeric(length(df))  
  for (i in seq_along(df)) {  
    output[i] <- fun(df[[i]])  
  }  
  output  
}
```

```
> col_summary(df, fun = median)  
> col_summary(df, fun = mean)  
> col_summary(df, fun = sd)
```





R Functions

**Let's practice!**



R Functions

# Introducing purrr



# Passing functions as arguments

```
> sapply(df, mean)
```

a	b	c	d
0.0643872	-0.1630165	-0.1057590	0.0406435

```
> col_summary(df, mean)
```

```
[1] 0.0643872 -0.1630165 -0.1057590 0.0406435
```

```
> library(purrr)
```

```
> map_dbl(df, mean)
```

a	b	c	d
0.0643872	-0.1630165	-0.1057590	0.0406435

# Every map function works the same way

```
map_dbl(.x, .f, ...)
```

1. Loop over a vector `.x`
2. Do something to each element `.f`
3. Return the results

# The map functions differ in their return type

There is one function for each type of vector:

- `map()` returns a list
- `map_dbl()` returns a double vector
- `map_lgl()` returns a logical vector
- `map_int()` returns a integer vector
- `map_chr()` returns a character vector

# Different types of vector input

```
map(.x, .f, ...)
```

**.x is always a vector**

```
> df <- data.frame(a = 1:10, b = 11:20)
> map(df, mean)
$a
[1] 5.5

$b
[1] 15.5
```

**Data frames, iterate over columns**

# Different types of vector input

```
> l <- list(a = 1:10, b = 11:20)
> map(l, mean)
$a
[1] 5.5

$b
[1] 15.5
```

**Lists, iterate over elements**

# Different types of vector input

```
> vec <- c(a = 1, b = 2)
> map(vec, mean)
$a
[1] 1

$b
[1] 2
```

**Vectors, iterate over elements**

# Advantages of the map functions in purrr

- Handy shortcuts for specifying `. f`
- More consistent than `sapply()`, `lapply()`, which makes them better for programming (Chapter 5)
- Takes much less time to solve iteration problems



R Functions

**Let's practice!**





R Functions

# Shortcuts for specifying `.f`

# Specifying `.f`

```
> map(df, summary)
```

## An existing function

```
> map(df, rescale01)
```

## An existing function you defined

```
> map(df, function(x) sum(is.na(x)))
```

## An anonymous function defined on the fly

```
> map(df, ~ sum(is.na(.)))
```

## An anonymous function defined using a formula shortcut

# Shortcuts when `.f` is `[[`

```
> list_of_results <- list(  
  list(a = 1, b = "A"),  
  list(a = 2, b = "C"),  
  list(a = 3, b = "D")  
)  
  
> map_dbl(list_of_results, function(x) x[["a"]]) An anonymous function  
[1] 1 2 3  
  
> map_dbl(list_of_results, "a") Shortcut: string subsetting  
[1] 1 2 3  
  
> map_dbl(list_of_results, 1) Shortcut: integer subsetting  
[1] 1 2 3
```

# A list of data frames

```
> cyl <- split(mtcars, mtcars$cyl)
> str(cyl)
```

List of 3

\$ 4:'data.frame': 11 obs. of 11 variables:

```
..$ mpg : num [1:11] 22.8 24.4 22.8 32.4 30.4 33.9 21.5 ...
..$ cyl : num [1:11] 4 4 4 4 4 4 4 4 4 4 4 ...
```

...

\$ 6:'data.frame': 7 obs. of 11 variables:

```
..$ mpg : num [1:7] 21 21 21.4 18.1 19.2 17.8 19.7 ...
..$ cyl : num [1:7] 6 6 6 6 6 6 6 ...
```

...

\$ 8:'data.frame': 14 obs. of 11 variables:

```
..$ mpg : num [1:14] 18.7 14.3 16.4 17.3 15.2 10.4 10.4 14.7 ...
..$ cyl : num [1:14] 8 8 8 8 8 8 8 8 8 8 8 ...
```

**Split the data frame `mtcars` based on the unique values in the `cyl` column**

# A list of data frames

```
> cyl[[1]]
```

	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2

# Goal

- Fit regression to each of the data frames in `cyl`
- Quantify relationship between `mpg` and `wt`

```
# Slopes for regressions on mpg on weight for each cylinder class
      4          6          8
-5.647025 -2.780106 -2.192438
```



R Functions

**Let's practice!**