# More data wrangling with dplyr

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# Data Wrangling Pipelines

### Toy Data

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50

```
dat <- data.frame(
   name = c('Anakin', 'Padme', 'Luke', 'Leia'),
   gender = c('male', 'female', 'male', 'female'),
   height = c(1.88, 1.65, 1.72, 1.50)
)</pre>
```

# Function calls in dplyr

## dplyr functional calls

An "ugly" side of dplyr is that if you want to do many operations at once, it does not lead to particularly elegant code.

You either have to do computations, step-by-step, with separate commands...

Or you have to wrap several function calls inside each other (making your code hard to read)

#### output

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



gender	avg	sd
male	1.8	0.113
female	1.58	0.106

Example: For each gender category, get the average and standard deviation of height, arranging output by average in descending order.

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



gender	avg	sd
male	1.8	0.113
female	1.58	0.106

Step-by-step computations

```
dat1 = group_by(dat, gender)
dat2 = summarise(dat1,
    avg = mean(height), sd = sd(height))
dat3 = arrange(dat2, desc(avg))
dat3
```

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



gender	avg	sd
male	1.8	0.113
female	1.58	0.106

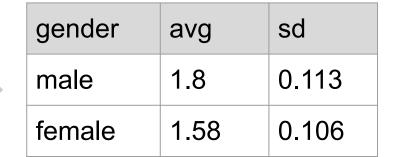
Function calls inside each other

```
arrange(
   summarise(group_by(dat, gender),
   avg = mean(height),
   sd = sd(height)),
   desc(avg))
```

# Pipe Operators | > or %>%

dat dat3

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



Pipeline of commands: easier to write and understand

# Pipe operators



Base R



magrittr

## Pipe operators

A pipe operator lets you write:

as:

$$x \mid > f(y)$$

or equivalently as:

$$x \%>\% f(y)$$

### Example

```
x = c(2, 4, 6, NA)
# without the pipe
mean(x, na.rm = TRUE)
# with the pipe
x |> mean(na.rm = TRUE)
```

```
set.seed(12345)
n = 10
x1 = runif(n, min = -3, max = 3)
x2 = round(x1, 2)
                                       no pipe
x3 = abs(x2)
x4 = sum(x3)
x4
set.seed(12345)
10 |>
  runif(min = -3, max = 3) |>
  round(2) |>
  abs() |>
  sum()
```

## Pipe Operators: %>% and |>

**Pipe operators**, denoted as %>% and also as |>, allow you to write function calls in a more human-readable way.

These operators are heavily used among the ecosystem of "tidyverse" packages, and they are becoming more common in traditional R code.

Technically speaking, %>% is known as the "magrittr" pipe operator (from its homonym package, introduced in 2014).

In turn, > is the base R pipe operator (introduced in May 2021 with R version 4.1.0.)

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



name	gender	height
Padme	female	1.65
Leia	female	1.50

# is equivalent to
dat |> filter(gender == "female")

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



name	gender
Anakin	male
Padme	female
Luke	male
Leia	female

select(dat, name:gender)

# is equivalent to
dat |> select(name:gender)

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



name	gender	height
Anakin	male	1.88
Luke	male	1.72
Padme	female	1.65
Leia	female	1.50

arrange(dat, desc(height))

# is equivalent to
dat |> arrange(desc(height))

dat

name	gender	height	
Anakin	male	1.88	
Padme	female	1.65	
Luke	male	1.72	
Leia	female	1.50	

Example: number of non-male individuals with heights greater than 1.40 meters

name	gender height	
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50

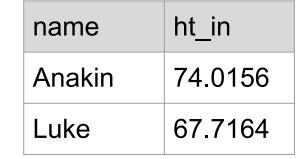


```
dat |>
  filter(gender != "male") |>
  filter(height > 1.4) |>
  summarise(n())
```

name	gender	height	
Anakin	male	1.88	
Padme	female	1.65	
Luke	male	1.72	
Leia	female	1.50	

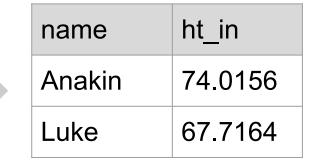
Example: convert height into inches, and display name and height (inches) of male individuals, arranging content by height in descending order.

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



```
dat |>
  mutate(ht_in = height * 39.37) |>
  filter(gender == "male") |>
  select(name, ht_in) |>
  arrange(desc(ht in))
```

name	gender	height
Anakin	male	1.88
Padme	female	1.65
Luke	male	1.72
Leia	female	1.50



Equivalent pipeline

```
dat |>
  filter(gender == "male") |>
  mutate(ht_in = height * 39.37) |>
  select(name, ht_in) |>
  arrange(desc(ht_in))
```

# Merging Tables with joins()

# **Data Tables**

#### tbl1

id	year coffee	
Luke	1	no
Leia	3	yes
Han	4	yes

#### tbl2

id	gpa	lunch	
Padme	3.9	pizza	
Leia	4.0	tacos	
Luke	3.7	burrito	
Obi-Wan	3.8	pad thai	

#### **Data Tables**

```
tbl1 <- data.frame(
 id = c('Luke', 'Leia', 'Han'),
 year = c(1, 3, 4),
 coffee = c('no', 'yes', 'yes')
tbl2 <- data.frame(
 id = c('Padme', 'Leia', 'Luke', 'Obi-Wan'),
 qpa = c(3.9, 4.0, 3.7, 3.8),
 lunch = c('pizza', 'tacos', 'burrito', 'pad thai')
```

tbl1

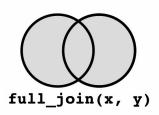
id	year coffee	
Luke	1	no
Leia	3	yes
Han	4	yes

#### tbl2

id	gpa	lunch	
Padme	3.9	pizza	
Leia	4.0	tacos	
Luke	3.7	burrito	
Obi-Wan	3.8	pad thai	

# keeps all observations in tbl1 and tbl2
full\_join(tbl1, tbl2, by = "id")

id	year	coffee	gpa	lunch
Luke	1	no	3.7	burrito
Leia	3	yes	4.0	tacos
Han	4	yes	NA	NA
Padme	NA	NA	3.9	pizza
Obi-Wan	NA	NA	3.8	pad thai



tbl1

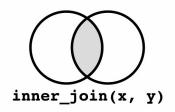
id	year	coffee
Luke	1	no
Leia	3	yes
Han	4	yes

tbl2

id	gpa	lunch
Padme	3.9	pizza
Leia	4.0	tacos
Luke	3.7	burrito
Obi-Wan	3.8	pad thai

# keeps obs in tbl1 that have matching key in tbl2
inner\_join(tbl1, tbl2, by = "id")

id	year	coffee	gpa	lunch
Luke	1	no	3.7	burrito
Leia	3	yes	4.0	tacos



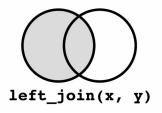
tbl1 tbl2

id	year	coffee
Luke	1	no
Leia	3	yes
Han	4	yes

id	gpa	lunch
Padme	3.9	pizza
Leia	4.0	tacos
Luke	3.7	burrito
Obi-Wan	3.8	pad thai

# keeps all observations tbl1
left\_join(tbl1, tbl2, by = "id")

id	year	coffee	gpa	lunch
Luke	1	no	3.7	burrito
Leia	3	yes	4.0	tacos
Han	4	yes	NA	NA



tbl1

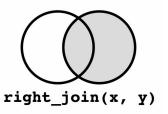
id	year	coffee
Luke	1	no
Leia	3	yes
Han	4	yes

#### tbl2

id	gpa	lunch
Padme	3.9	pizza
Leia	4.0	tacos
Luke	3.7	burrito
Obi-Wan	3.8	pad thai

# keeps all observations in tbl2
right\_join(tbl1, tbl2, by = "id")

id	year	coffee	gpa	lunch
Luke	1	no	3.7	burrito
Leia	3	yes	4.0	tacos
Padme	NA	NA	3.9	pizza
Obi-Wan	NA	NA	3.8	pad thai



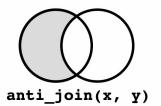
tbl1 tbl2

id	year	coffee
Luke	1	no
Leia	3	yes
Han	4	yes

id	gpa	lunch
Padme	3.9	pizza
Leia	4.0	tacos
Luke	3.7	burrito
Obi-Wan	3.8	pad thai

# return rows from tbl1 without a match in tbl2
anti\_join(tbl1, tbl2, by = "id")

id	year	coffee
Han	4	yes



tbl1

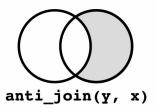
id	year	coffee
Luke	1	no
Leia	3	yes
Han	4	yes

tb	12
----	----

id	gpa	lunch	
Padme	3.9	pizza	
Leia	4.0	tacos	
Luke	3.7	burrito	
Obi-Wan	3.8	pad thai	

# return rows from tbl2 without a match in tbl1
anti\_join(tbl2, tbl1, by = "id")

id	gpa	lunch
Padme	3.9	pizza
Obi-Wan	3.8	pad thai



tbl1 tbl2

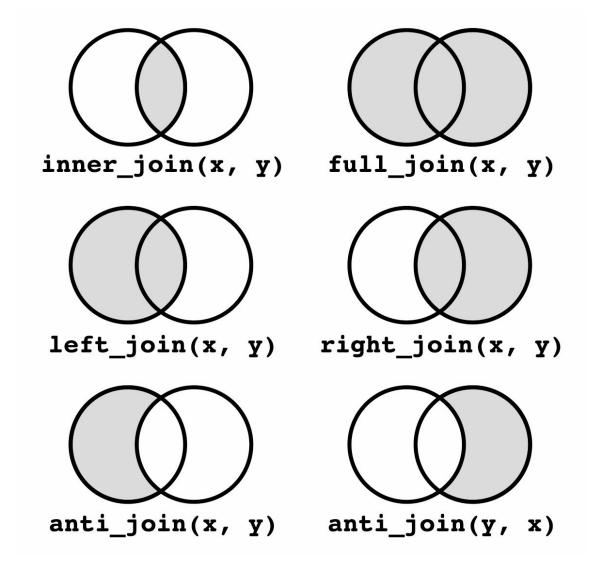
id	year coffee	
Luke	1	no
Leia	3	yes
Han	4	yes

id	gpa	lunch	
Padme	3.9	pizza	
Leia	4.0	tacos	
Luke	3.7	burrito	
Obi-Wan	3.8	pad thai	

# return rows from tbl1 with a match in tbl2
semi join(tbl1, tbl2, by = "id")

id	year	coffee
Luke	1	no
Leia	3	yes

### **Joins**



# What if tables had keys with different names?

# Tables with different key names

tbl1

id1	year coffee		
Luke	1	no	
Leia	3	yes	
Han	4	yes	

tbl2

id2	gpa	lunch	
Padme	3.9	pizza	
Leia	4.0	tacos	
Luke	3.7	burrito	
Obi-Wan	3.8	pad thai	

#### **Data Tables**

```
tbl1 <- data.frame(
 id1 = c('Luke', 'Leia', 'Han'),
 year = c(1, 3, 4),
 coffee = c('no', 'yes', 'yes')
tbl2 <- data.frame(
 id2 = c('Padme', 'Leia', 'Luke', 'Obi-Wan'),
 qpa = c(3.9, 4.0, 3.7, 3.8),
 lunch = c('pizza', 'tacos', 'burrito', 'pad thai')
```

tbl1 tbl2

id1	year coffee		
Luke	1	no	
Leia	3	yes	
Han	4	yes	

id2	gpa	lunch	
Padme	3.9	pizza	
Leia	4.0	tacos	
Luke	3.7	burrito	
Obi-Wan	3.8	pad thai	

# keeps all observations in tbl1 and tbl2
full\_join(tbl1, tbl2, join\_by("id1" == "id2"))

id1	year	coffee	gpa	lunch
Luke	1	no	3.7	burrito
Leia	3	yes	4.0	tacos
Han	4	yes	NA	NA
Padme	NA	NA	3.9	pizza
Obi-Wan	NA	NA	3.8	pad thai

tbl1 tbl2

id1	year	coffee	
Luke	1	no	
Leia	3	yes	
Han	4	yes	

id2	gpa	lunch	
Padme	3.9	pizza	
Leia	4.0	tacos	
Luke	3.7	burrito	
Obi-Wan	3.8	pad thai	

id1	year	coffee	gpa	lunch
Luke	1	no	3.7	burrito
Leia	3	yes	4.0	tacos
Han	4	yes	NA	NA
Padme	NA	NA	3.9	pizza
Obi-Wan	NA	NA	3.8	pad thai