MFE R Programming Workshop Week 6

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Introduction

Questions

Any questions before we start?

Overview

- **>** %>%
- ▶ tidyr
- ► dplyr



The Pipe Operator %>%

- ► The magnittr package provides a pipe operator.
- ► See vignette("magrittr").
- Basic piping:
 - x %>% f is equivalent to f(x)
 - x %>% f(y) is equivalent to f(x, y)
 - x %>% f %>% g %>% h is equivalent to h(g(f(x)))
- ▶ The argument placeholder:
 - x %>% f(y, .) is equivalent to f(y, x)
 - \rightarrow x %>% f(y, z = .) is equivalent to f(y, z = x)

Expose the variables with %\$%

► The %\$% allows variable names (e.g. column names) to be used in a function.

```
library(magrittr)
iris %>%
subset(Sepal.Length > mean(Sepal.Length)) %$%
cor(Sepal.Length, Sepal.Width)
```

```
## [1] 0.3361992
```

Compound assignment pipe operations with %<>%

There is also a pipe operator which can be used as shorthand notation in situations where the left-hand side is being "overwritten":

```
iris$Sepal.Length <-
iris$Sepal.Length %>%
sqrt()
```

Use the %<>% operator to avoid the repetition:

```
iris$Sepal.Length %<>% sqrt
```

► This operator works exactly like %>%, except the pipeline assigns the result rather than returning it.

tidyr

Hadley Wickham

- ► Hadley Wickham is practically famous in the R world
- ► He's developed a very large number of useful packages, e.g. ggplot2 and lubridate.
- Today we will look at dplyr and tidyr.
- Tidy data is data that's easy to work with: it's easy to munge (with dplyr), visualise (with ggplot2 or ggvis) and model (with R's hundreds of modelling packages).
- ▶ The two most important properties of tidy data are:
 - ► Each column is a variable.
 - Each row is an observation.

Sample data

tidyr is a member of the core tidyverse.

```
library(tidyverse)
set.seed(1234)
stocks <- data.frame(
time = as.Date('2009-01-01') + 0:4,
X = rnorm(5, 0, 1),
Y = rnorm(5, 0, 2),
Z = rnorm(5, 0, 4)
)
head(stocks, n = 4)</pre>
```

```
## time X Y Z
## 1 2009-01-01 -1.2070657 1.012112 -1.9087708
## 2 2009-01-02 0.2774292 -1.149480 -3.9935458
## 3 2009-01-03 1.0844412 -1.093264 -3.1050156
## 4 2009-01-04 -2.3456977 -1.128904 0.2578353
```

Bring columns together with gather()

```
stocksm <- stocks %>% gather(stock, price, -time)
stocksm
```

##		time	stock	price
##	1	2009-01-01	X	-1.2070657
##	2	2009-01-02	X	0.2774292
##	3	2009-01-03	X	1.0844412
##	4	2009-01-04	X	-2.3456977
##	5	2009-01-05	Х	0.4291247
##	6	2009-01-01	Y	1.0121118
##	7	2009-01-02	Y	-1.1494799
##	8	2009-01-03	Y	-1.0932637
##	9	2009-01-04	Y	-1.1289040
##	10	2009-01-05	Y	-1.7800757
##	11	2009-01-01	Z	-1.9087708
##	12	2009-01-02	Z	-3.9935458
##	13	2009-01-03	Z	-3.1050156
##	14	2009-01-04	Z	0.2578353

Split a column with spread()

```
stocksm %>% spread(stock, price)
```

```
##
                        χ
          time
## 1 2009-01-01 -1.2070657 1.012112 -1.9087708
## 2 2009-01-02 0.2774292 -1.149480 -3.9935458
## 3 2009-01-03 1.0844412 -1.093264 -3.1050156
## 4 2009-01-04 -2.3456977 -1.128904 0.2578353
## 5 2009-01-05 0.4291247 -1.780076 3.8379762
```

```
stocksm %>% spread(time, price)
```

##

```
stock 2009-01-01 2009-01-02 2009-01-03 2009-01-04 2009
        X -1.207066 0.2774292 1.084441 -2.3456977
## 1
                                                     0.4
        Y 1.012112 -1.1494799 -1.093264 -1.1289040 -1.
## 2
## 3
        Z -1.908771 -3.9935458 -3.105016 0.2578353 3.8
```

spread() and gather() are complements

```
df \leftarrow data.frame(x = c("a", "b"), y = c(3, 4),
                 z = c(5, 6)
df
## x y z
## 1 a 3 5
## 2 b 4 6
df %>% spread(x, y) %>% gather(x, y, a:b, na.rm = TRUE)
## z x y
## 1 5 a 3
## 4 6 b 4
```

There's much more

► As usual, read the vignette on the CRAN page

dplyr

Overview of dplyr

- dplyr provides a grammar of data manipulation.
 - ► A simple way to interact with data.
- We learn about:
 - ▶ tibble structure tbl
 - ► The pipe operator %>%
 - Using dplyr with databases
- ► The dplyr introduction vignette is a good resource.

dplyr and data.table

- ► See this post.
- Here are my thoughts:
 - ► For data less than 1 million rows, it is reported that there is not a significant speed difference between the two.
 - ▶ For large data that can fit in memory, use data.table.
 - For data than cannot fit in memory, you could use dplyr with a database backend.
- dtplyr is a package to use dplyr with data.table.
 - It is slower than just using data.table.

Data: nycflights13

- ► To explore the basic data manipulation verbs of dplyr, we'll start with the built in 'nycflights13} data frame
- ▶ This dataset contains all flights that departed from New York City in 2013

```
library(dplyr)
library(nycflights13)
```

head(flights,4)

```
# A tibble: 4 \times 19
##
##
     year month day dep_time sched_dep_time dep_delay
##
     <int> <int> <int>
                                                  <dbl>
                         <int>
                                        <int>
                                          515
## 1
     2013
                           517
## 2 2013
                           533
                                          529
                                          540
## 3 2013
                           542
## 4 2013
                           544
                                          545
     ... with 13 more variables: arr time <int>,
```

tbls (Tibbles)

##

7 ## 8

##

► A tbl will only display the data that will fit in your console.
-glimpse() is another nice way to look at the data

```
flights <- tbl_df(flights)
flights</pre>
```

A tibble: 336,776 × 19

2013

2013

2013

```
##
       year month
                      day dep_time sched_dep_time dep_delay
##
      <int> <int> <int>
                                                         <dbl>
                             <int>
                                              <int>
## 1
       2013
                                517
                                                515
## 2
       2013
                                533
                                                529
## 3
       2013
                                542
                                                540
## 4
       2013
                                544
                                                545
                                                            -1
## 5
       2013
                                554
                                                600
                                                            -6
       2013
                                554
                                                558
##
                                                            -4
```

555

557

557

600

600

600

-5

-3

1

Single Table Verbs

- dplyr aims to provide a function for each basic verb of data manipulation:
- select() (and rename())
 - returns a subset of the columns
- filter() (and slice())
 - returns a subset of the rows
- arrange() reorders rows
 - reorders the rows according to single or multiple variables
- distinct()
- mutate() (and transmute())
 - builds adds new columns from the data
- summarise() calculates summary statistics
 - which reduces each group to a single row by calculating aggregate measures
- sample_n() and sample_frac()

Tidy Data

- dplyr works best when variables are in columns and observations are in rows.
- ▶ You can use tidyr to help you create a tidy dataset.

Select Columns by Name with select()

select() allows you to rapidly zoom in on a useful subset using operations that usually only work on numeric variable positions:

```
# Select columns by name
select(flights, year, month, day)
```

```
## # A tibble: 336,776 \times 3
##
      year month
                  dav
##
     <int> <int> <int>
## 1
      2013
## 2 2013 1
## 3 2013 1
      2013 1
## 4
## 5
      2013
## 6
      2013
## 7
      2013
      2013
```

Select a Range of Columns with:

```
# Select all columns between year and day (inclusive) select(flights, year:day)
```

```
## # A tibble: 336,776 \times 3
##
      year month
                   day
## <int> <int> <int>
## 1
    2013
## 2 2013
## 3 2013 1
    2013 1
## 4
## 5
      2013
## 6
      2013
      2013
## 7
      2013
## 8
      2013
## 9
## 10
      2013
  # ... with 336,766 more rows
```

An Example of -(col1:col2)

557

557

558

8

9

#

10

Select all columns except those from year to day (inclus select(flights, -(year:day))

```
## # A tibble: 336,776 × 16
##
      dep_time sched_dep_time dep_delay arr_time
##
         <int>
                         <int>
                                    <dbl>
                                              <int>
## 1
           517
                            515
                                                830
## 2
           533
                            529
                                        4
                                                850
## 3
           542
                            540
                                                923
```

4 544 545 -1 1004

5 554 600 -6 812

6 554 558 740 -4 ## 7 555 600 -5 913

600

600

600

... with 336,766 more rows, and 12 more variables:

sched arr time (int) arr delay (dhl) carrier (chr)

709

838

753

-3

-3

-2

select Helper Functions

- dplyr comes with a set of helper functions that can help you select groups of variables inside a select() call:
- starts_with("X"): every name that starts with "X",
- ▶ ends_with("X"): every name that ends with "X",
- contains("X"): every name that contains "X",
- matches("X"): every name that matches "X", where "X" can be a regular expression,
- num_range("x", 1:5): the variables named x01, x02, x03, x04 and x05,
- one_of(x): every name that appears in x, which should be a character vector.

Add New Columns with mutate()

```
mutate(flights,
       gain = arr_delay - dep_delay,
       speed = distance / air_time * 60)
```

```
# A tibble: 336,776 × 21
##
      year month day dep_time sched_dep_time dep_delay
                                                    <4h1>
```

##	/III (/	TII (/)	LII U /	\111C/	\1II U /	\ab1>
## 1	2013	1	1	517	515	2
## 2	2013	1	1	533	529	4
## 3	2013	1	1	542	540	5

#	# 1	2013	1	1	517	515	2
#	# 2	2013	1	1	533	529	4
#	# 3	2013	1	1	542	540	2

##	2	2013	1	1	533	529	4
##	3	2013	1	1	542	540	2
##	4	2013	1	1	544	545	-1
##	5	2013	1	1	554	600	-6
##	6	2013	1	1	554	558	-4

## 3	2013	1	1	542	540	2
## 4	2013	1	1	544	545	-1
## 5	2013	1	1	554	600	-6
## 6	2013	1	1	554	558	-4
## 7	2013	1	1	555	600	-5
## 8	2013	1	1	557	600	-3
## 9	2013	1	1	557	600	-3

2013 558 600 ## 10 with 336.766 more rows, and 15 more variables $^{27/44}$

If you only want to keep the new variables, use transmute()

```
## # A tibble: 336,776 \times 2
##
      gain gain_per_hour
     <dbl>
                  <dbl>
##
## 1
             2.378855
    16
## 2
             4.229075
    31
          11.625000
## 3
## 4
    -17
          -5.573770
## 5
    -19
          -9.827586
## 6
    16
              6.400000
               9.113924
## 7
    24
          -12.452830
## 8
      -11
```

Filter rows with filter()

- ▶ filter() allows you to select a subset of rows in a data frame.
- ▶ The first argument is the name of the data frame.
- ► The second and subsequent arguments are the expressions that filter the data frame
- ► Select all flights on January 1st with:

A tibble: 842 × 19

vear month

2013

2013

2013

##

5

##

##

```
filter(flights, month == 1, day == 1)
```

```
##
      <int> <int> <int>
                              <int>
                                                           <dbl>
                                               <int>
## 1
       2013
                                517
                                                 515
                                                               2
## 2
       2013
                                533
                                                 529
                         1
## 3
       2013
                                542
                                                 540
       2013
                                544
                                                 545
##
                                                              -1
```

554

554

555

1

day dep_time sched_dep_time dep_delay

600

558

600

-6

Select rows by position

► To select rows by position, use slice()

```
slice(flights, 1:10)
```

```
# A tibble: 10 × 19
##
                      day dep_time sched_dep_time dep_delay
       vear month
##
      <int> <int> <int>
                              <int>
                                              <int>
                                                         <dbl>
##
       2013
                                517
                                                 515
                                                              2
## 2
       2013
                                533
                                                 529
                                                              4
## 3
       2013
                                542
                                                540
                                544
##
   4
       2013
                                                545
                                                             -1
## 5
       2013
                                554
                                                600
                                                             -6
## 6
       2013
                                554
                                                 558
                                                             -4
## 7
       2013
                                555
                                                600
                                                             -5
       2013
                                557
                                                600
                                                             -3
## 8
       2013
                        1
                                557
                                                600
                                                             -3
##
##
   10
       2013
                                558
                                                600
                                                             -2
     ... with 13 more variables: arr time <int>,
```

Arrange rows with arrange()

2013

2013

2013

2013

7 ## 8

##

10

arrange() works similarly to filter() except that instead of filtering or selecting rows, it reorders them.

```
arrange(flights, year, month, day)
```

A tibble: 336,776 × 19

```
##
       vear month
                     day dep_time sched_dep_time dep_delay
##
      <int> <int> <int>
                             <int>
                                              <int>
                                                         <dbl>
## 1
       2013
                               517
                                                515
## 2
       2013
                               533
                                                529
                               542
## 3
       2013
                                                540
## 4
       2013
                               544
                                                545
                                                            -1
## 5
       2013
                               554
                                                600
                                                            -6
## 6
       2013
                               554
                                                558
                                                            -4
```

1

555

557

557

558

600

600

600

600

-5

-3

-3

Use desc() to order a column in descending order

arrange(flights, desc(arr_delay))

A tibble: 336,776 × 19

7

8

#

#

```
year month day dep time sched dep time dep delay
##
##
      <int> <int> <int>
                            <int>
                                             <int>
                                                       <dbl>
## 1
       2013
                 1
                       9
                               641
                                               900
                                                        1301
## 2
       2013
                 6
                      15
                              1432
                                              1935
                                                        1137
## 3
       2013
                      10
                              1121
                                              1635
                                                        1126
                 9
## 4
       2013
                      20
                              1139
                                              1845
                                                        1014
## 5
       2013
                      22
                             845
                                              1600
                                                        1005
## 6
       2013
                4
                      10
                              1100
                                              1900
                                                         960
```

arr time <int>, sched arr time <int>, arr delay <db

10

carrier (chr) flight (int) tailnum (chr)

... with 336,766 more rows, and 13 more variables:

9

You can rename variables with rename()

rename(flights, tail_num = tailnum)

#

```
## # A tibble: 336,776 × 19
       year month day dep time sched dep time dep delay
##
      <int> <int> <int>
                           <int>
                                           <int>
                                                     <dbl>
##
## 1
       2013
                              517
                                             515
                                                         2
## 2
       2013
                              533
                                             529
## 3
       2013
                              542
                                             540
## 4
       2013
                              544
                                             545
                                                        -1
## 5
       2013
                              554
                                             600
                                                        -6
## 6
       2013
                              554
                                             558
                                                        -4
## 7
       2013
                              555
                                             600
                                                        -5
## 8
       2013
                              557
                                             600
                                                        -3
```

... with 336,766 more rows, and 13 more variables:

9 2013 557 600 -3 ## 10 2013 558 600 -2

arr time <int>, sched arr time <int>, arr delay <db carrier (chr) flight (int) tail num (chr)

Extract distinct (unique) rows

- ► A common use of select() is to find the values of a set of variables.
- ► This is particularly useful in conjunction with the distinct() verb

```
distinct(select(flights, tailnum))
```

```
## # A tibble: 4,044 × 1
##
      tailnum
##
        <chr>
## 1 N14228
## 2 N24211
## 3
       N619AA
## 4
       N804.JB
       N668DN
## 5
## 6
       N39463
## 7
       N516.JB
       N829AS
##
```

Summarise values with summarise()

- The last verb is summarise(). It collapses a data frame to a single row.
- You can use any function you like in summarise() so long as the function can take a vector of data and return a single number.

```
## # A tibble: 1 × 1
## delay
## <dbl>
## 1 12.63907
```

dplyr aggregate functions

- dplyr provides several helpful aggregate functions of its own, in addition to the ones that are already defined in R. These include:
 - first(x) The first element of vector x.
 - last(x) The last element of vector x.
 - nth(x, n) The nth element of vector x.
 - n() The number of rows in the data.frame or group of observations that summarise() describes.
 - n_distinct(x) The number of unique values in vector x.

Chaining

- ► The dplyr API is functional function calls don't have side-effects.
- You must always save their results. UGLY
- ► To get around this problem, dplyr provides the %>% operator
- \triangleright x %>% f(y) turns into f(x, y)

```
flights %>%
group_by(year, month, day) %>%
select(arr_delay, dep_delay) %>%
summarise(arr = mean(arr_delay, na.rm = TRUE),
dep = mean(dep_delay, na.rm = TRUE)) %>%
filter(arr > 30 | dep > 30)
```

Adding missing grouping variables: `year`, `month`, `day

Source: local data frame [49 x 5]
Groups: year, month [11]
##

Commonalities

- ▶ The syntax and function of all these verbs are very similar:
- ▶ The first argument is a data frame.
- ► The subsequent arguments describe what to do with the data frame.
- The result is a new data frame
- ► Together these properties make it easy to chain together multiple simple steps to achieve a complex result.

Grouped operations

- ► These verbs are useful on their own, but they become really powerful when you apply them to groups of observations
- ▶ In dplyr, you do this by with the group_by() function.
- ▶ It breaks down a dataset into specified groups of rows.

Grouped operations (cont.)

- Grouping affects the verbs as follows:
- grouped select() is the same as ungrouped select(), except that grouping variables are always retained.
- grouped arrange() orders first by the grouping variables
- mutate() and filter() are most useful in conjunction with window functions (like rank(), or min(x) = x=). They are described in detail in vignette("window-functions").
- sample_n() and sample_frac() sample the specified number/fraction of rows in each group.
- slice() extracts rows within each group.
- summarise() is powerful and easy to understand, as described in more detail below.

group_by Example

► For example, we could use these to find the number of planes and the number of flights that go to each possible destination:

```
## # A tibble: 105 × 3
##
      dest planes flights
##
     <chr> <int>
                    <int>
## 1
       ABQ
              108
                      254
## 2
       ACK 58
                      265
## 3
       ALB
           172
                      439
       ANC
## 4
                6
                        8
       ATT.
## 5
             1180
                    17215
##
       AUS
              993
                     2439
       AVT.
              159
                      275
##
```

Multiple table verbs

- dplyr implements the four most useful SQL joins:
- ▶ inner_join(x, y): matching x + y
- ▶ left_join(x, y): all x + matching y
- semi_join(x, y): all x with match in y
- ▶ anti_join(x, y): all x without match in y
- And provides methods for:
- ▶ intersect(x, y): all rows in both x and y
- ▶ union(x, y): rows in either x or y
- setdiff(x, y): rows in x, but not y

Joins from dplyr Map to SQL

- ▶ inner_join(x, y)
 - ► SELECT * FROM x JOIN y ON x.a = y.a
- ▶ left_join(x, y)
 - ► SELECT * FROM x LEFT JOIN y ON x.a = y.a
- ▶ right_join(x, y)
 - ► SELECT * FROM x RIGHT JOIN y ON x.a = y.a
- ► full_join(x, y)
 - ► SELECT * FROM x FULL JOIN y ON x.a = y.a
- semi_join(x, y)
 - ► SELECT * FROM x WHERE EXISTS (SELECT 1 FROM y WHERE x.a = y.a)
- ▶ anti_join(x, y)
 - ► SELECT * FROM x WHERE NOT EXISTS (SELECT 1 FROM y WHERE x.a = y.a)

Lab 3

▶ Let's redo Lab 3 with dplyr.