Multi-Page dplyr

Andy Grogan-Kaylor

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1 Background

`dplyr` is a very powerful R library for managing and processing data.[1]

While `dplyr` is very powerful, learning to use `dplyr` can be very confusing. This guide aims to present some of the most common `dplyr` functions and commands in the form of a brief cheatsheet.

[1] The origins of the name `dplyr` seem somewhat obscure, but I sometimes think of this package as the *data plyers*.

```
library(dplyr)
```

2 Simulated Data

year	x	y	z
2016	NA	Group A	103
2018	30.17	Group C	110.9
2017	38.84	Group B	105.9

year	x	y	z
2017	50.45	Group B	100.4
2015	59.04	Group A	104.1

3 Piping

Pipes `%>%` connect pieces of a command e.g. *data* to *data wrangling* to a *graph command*.

4 Aggregate Data: `group_by()` & `summarise()`

```
mynewdata <- mydata %>%
  group_by(year) %>% # group by y
  summarise(mean_x = mean(x), # mean of x
            n = n()) # count up
```

year	mean_x	n
2015	59.04	1
2016	NA	1
2017	44.65	2
2018	30.17	1

5 Select A Subset of Variables: `select()`

```
mynewdata <- mydata %>% select(x, y)  # select only x and y
```

x	y
NA	Group A
30.17	Group C
38.84	Group B
50.45	Group B
59.04	Group A

6 Filter A Subset of Rows: `filter()`

```
mynewdata <- mydata %>% filter(year > 2010)  # filter on year
```

year	x	y	z
2016	NA	Group A	103
2018	30.17	Group C	110.9
2017	38.84	Group B	105.9
2017	50.45	Group B	100.4
2015	59.04	Group A	104.1

7 Create New Variables: `mutate()`

```
mynewdata <- mydata %>% mutate(myscale = x + z)  # create a new variable e.g. a scale
```

year	x	y	z	myscale
2016	NA	Group A	103	NA
2018	30.17	Group C	110.9	141.1
2017	38.84	Group B	105.9	144.7
2017	50.45	Group B	100.4	150.9
2015	59.04	Group A	104.1	163.1

8 Recode Variables: `mutate()`

8.1 Continuous Into Categorical: `mutate()` & `cut()`

```
mynewdata <- mydata %>%
  mutate(zcategorical = cut(z, # cut at breaks
                            breaks=c(-Inf, 100, Inf),
                labels = c("low", "high")))
```

year	x	y	z	zcategorical
2016	NA	Group A	103	high
2018	30.17	Group C	110.9	high
2017	38.84	Group B	105.9	high
2017	50.45	Group B	100.4	high
2015	59.04	Group A	104.1	high

8.2 Categorical Into Categorical: `mutate()` & `recode()`

```
mynewdata <- mydata %>%
  mutate(yrecoded = dplyr::recode(y, # recode values
                          "Group A" = "Red Group",
                          "Group B" = "Blue Group",
```

```
                    .default = "Other"))
```

year	x	y	z	yrecoded
2016	NA	Group A	103	Red Group
2018	30.17	Group C	110.9	Other
2017	38.84	Group B	105.9	Blue Group
2017	50.45	Group B	100.4	Blue Group
2015	59.04	Group A	104.1	Red Group

9 Rename Variables: `rename()`

```
newdata <- mydata %>%
  rename(age = x, # rename
         mental_health = z)
```

year	age	y	mental_health
2016	NA	Group A	103
2018	30.17	Group C	110.9
2017	38.84	Group B	105.9
2017	50.45	Group B	100.4
2015	59.04	Group A	104.1

10 Drop Missing Values: `filter()`

```
newdata <- mydata %>% filter(!is.na(x))  # filter by x is not missing
```

year	x	y	z
2018	30.17	Group C	110.9
2017	38.84	Group B	105.9
2017	50.45	Group B	100.4
2015	59.04	Group A	104.1

11 Random Sample

```
newdata <- mydata %>% sample_frac(0.5)  # fraction of data to sample
```

year	x	y	z
2018	30.17	Group C	110.9

year	x	y	z
2015	59.04	Group A	104.1

12 Connecting To Other Packages Like `ggplot`

Notice how, in the code below, I never actually create the new data set `mynewdata`. I simply pipe `mydata` into a `dplyr` command, and pipe the result directly to `ggplot2`.

```
library(ggplot2)

mydata %>% # my data
  mutate(myscale = x + z) %>% # dplyr command to make new variable
  ggplot(aes(x = year, # the rest is ggplot
             y = myscale)) +
  geom_point() + # points
  geom_smooth(se = FALSE) + # smoother without confidence interval
  labs(title = "My Scale By Year") + # labels
  theme(axis.text.x = element_text(size = 10, # tweak theme
                                   angle = 90))
```