Multi-Page dplyr

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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
2017	NA	Group A	98.99
2016	25.96	Group C	84.64
2017	38.09	Group A	117

year	x	y	z
2015	67.66	Group B	105.3
2015	35.09	Group A	116.6

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	51.38	2
2016	25.96	1
2017	NA	2

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

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

x	y
NA	Group A
25.96	Group C
38.09	Group A
67.66	Group B
35.09	Group A

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

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

year	x	y	z
2017	NA	Group A	98.99
2016	25.96	Group C	84.64
2017	38.09	Group A	117
2015	67.66	Group B	105.3
2015	35.09	Group A	116.6

7   Create New Variables: `mutate()`

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

year	x	y	z	myscale
2017	NA	Group A	98.99	NA
2016	25.96	Group C	84.64	110.6
2017	38.09	Group A	117	155.1
2015	67.66	Group B	105.3	173
2015	35.09	Group A	116.6	151.7

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
2017	NA	Group A	98.99	low
2016	25.96	Group C	84.64	low
2017	38.09	Group A	117	high
2015	67.66	Group B	105.3	high
2015	35.09	Group A	116.6	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
2017	NA	Group A	98.99	Red Group
2016	25.96	Group C	84.64	Other
2017	38.09	Group A	117	Red Group
2015	67.66	Group B	105.3	Blue Group
2015	35.09	Group A	116.6	Red Group

9   Rename Variables: `rename()`

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

year	age	y	mental_health
2017	NA	Group A	98.99
2016	25.96	Group C	84.64
2017	38.09	Group A	117
2015	67.66	Group B	105.3
2015	35.09	Group A	116.6

10   Drop Missing Values: `filter()`

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

year	x	y	z
2016	25.96	Group C	84.64
2017	38.09	Group A	117
2015	67.66	Group B	105.3
2015	35.09	Group A	116.6

11   Random Sample

```
newdata <- mydata %>%
  sample_frac(.5) # fraction of data to sample
```

year	x	y	z
2017	NA	Group A	98.99
2015	67.66	Group B	105.3

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
  filter(y != "Group C") %>% # filter on values of y
  ggplot(aes(x = year, # the rest is ggplot
             y = myscale)) +
  geom_point() + # points
  geom_smooth(se = FALSE, # smoother without confidence interval
              method = "lm") + # linear smoother
  labs(title = "My Scale By Year") + # labels
  theme(axis.text.x = element_text(size = 10, # tweak theme
                                   angle = 90))
```