# Multi-Page dplyr

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#### 2021-10-02

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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.

library(dplyr)

#### 2 Simulated Data

<sup>&</sup>lt;sup>1</sup>The origins of the name dplyr seem somewhat obscure, but I sometimes think of this package as the *data plyers*.

year	Х	у	Z
2019	NA	Group B	90
2020	50	Group C	90
2020	NA	Group A	100
2021	70	Group A	110
2021	80	Group B	120

## 3 Piping

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

dplyr commands will often look something like the outline below.

```
mydata %>%
  data_wrangling %>%
  more_data_wrangling %>%
  graph_command
```

## 4 Aggregate Data: group\_by() & summarise()

Notice how when aggregating data, we need to be explicit about whether we are removing NA values.

year	mean_x	n
2019	NA	1
2020	50	2
2021	75	2

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

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

х	У
NA	Group B
50	Group C
NA	Group A
70	Group A
80	Group B

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

```
mynewdata <- mydata %>%
  filter(year > 2020) # filter on year
```

year	х	у	Z
2021	70	Group A	110
2021	80	Group B	120

#### 7 Create New Variables: mutate()

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

year	х	у	Z	myscale
2019	NA	Group B	90	NA
2020	50	Group C	90	140
2020	NA	Group A	100	NA
2021	70	Group A	110	180
2021	80	Group B	120	200

### 8 Recode Variables: mutate()

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

year	Х	у	Z	zcategorical
2019	NA	Group B	90	low
2020	50	Group C	90	low
2020	NA	Group A	100	low
2021	70	Group A	110	high
2021	80	Group B	120	high

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

year	x	у	Z	yrecoded
2019	NA	Group B	90	Blue Group
2020	50	Group C	90	Other
2020	NA	Group A	100	Red Group
2021	70	Group A	110	Red Group
2021	80	Group B	120	Blue Group

### 9 Rename Variables: rename()

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

year	age	У	mental_health
2019	NA	Group B	90
2020	50	Group C	90
2020	NA	Group A	100
2021	70	Group A	110
2021	80	Group B	120

## 10 Drop Missing Values: filter()

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

year	x	у	Z
2020	50	Group C	90
2021	70	Group A	110
2021	80	Group B	120

## 11 Random Sample

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

year	X	у	Z
2021	70	Group A	110
2019	NA	Group B	90

## 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)
```

