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# STA130 W3

Monday, October 28, 2019 11:39 PM

#### Tidy data

- 1. Each variable must have its **own column**.
- 2. Each observation must have its **own row**.
- 3. Each value must have its own cell.

#### A general rule of thumb:

It is easier to describe functional relationships **between variables** (e.g., z is a linear combination of x and y, density is the ratio of weight to volume) than **between rows**.

It is easier to make **comparisons** between groups of **observations** (e.g., average of group A vs. average of group B) than between **groups of columns**.

### Not tidy:

Colour	N
Brown	97
Green / blue / gray	9
Other	23

This is tidy!	Summary
head(eye_data, n=5) ## # A tibble: 5 x 2 ## ID eye_colour ## <int> <chr> ## 1 1 Brown ## 2 2 Brown</chr></int>	eye_data %>% group_by(eye_colour) %>% summarise(n=n()) ## # A tibble: 3 x 2 ## eye_colour n ## <chr></chr>
## 3 3 Brown	## 2 Green/blue/gray 9
## 4 4 Brown ## 5 5 Brown	## 3 Other 23

**Data wrangling** allows us to transform data frames to make them more useful to answer interesting questions.

The *ggplot* library implements a grammar of graphics.
Similarly the *dplyr* library presents a grammar for data wrangling.

The pipe operator (%>%) is used to perform an action on a dataframe.

Glimpse the college\_recent\_grads data frame: college\_recent\_grads %>% glimpse()

# Select variables/columns using select():

Focus only on a few **variables** in the frame: major, major\_category, # of male graduates (men), # of female graduates (women), and the median salary (median)

We use the  ${\bf select()}$  function from  ${\bf dplyr}$  to extract a dataframe with only these  ${\bf variables}$ 

college\_recent\_grads %>%
 select(major, major\_category, men, women, median)

## Select observations/rows using filter():

Extract only **observations** degrees in Computer Science and Mathematics

```
college_recent_grads %>%
    filter(major_category == "Computers & Is TRUE
    Mathematics")

*Only keep rows where it is TRUE
!= : filtered out C & M
```

### Combining select() and filter():

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```
Extract certain subsets of the data, and save the new data frame as an R
object by giving it a name.
                             *give the new data frame a name to save it
CS_Math_grads <- college_recent_grads %>%
 select(major, major_category, men, women, median) %>%
 filter(major_category == "Computers & Mathematics")
```

We can use Logicals to write conditions on the variables in filter() to extract only the observations/rows where the condition is true.

#### Create new variables using mutate():

For each of the majors where the median earnings is at least \$60,000, what percentage of the graduates are women?

```
college_recent_grads %>%
 filter(median >= 60000) %>%
 select(major, men, women) %>%
 mutate(total = men + women,
                                        total: variable for total # of
    pct_female = round((women /
                                       graduates
total)*100, 2))
                                        pct female: % of female
                                       graduates
```

\*calculate the percentage of women grads for each program

### Sort observations based on new or existing variables using arrange():

```
college_recent_grads %>%
 filter(median >= 60000) %>%
 select(major, men, women) %>%
 mutate(total = men + women,
    pct_female = round((women / total)*100, 2)) %>%
arrange(pct_female) Sort values in a column, from smallest to largest
```

If sorting words, it will be in alphabetical order arrange(desc(pct\_female)) sorts in descending order

#### Create new variables from existing variables using mutate() and ifelse():

Create a categorical variable to identify majors with approximately equal numbers of male and female graduates: majors with between 45% and 55% female graduates.

Use ifelse() in a mutate() statement.

The format:

ifelse(test condition (logical), yes, no)

```
percent <- c(40, 47, 55, 58);
gender balance <- ifelse(percent >= 45 & percent <= 55, yes = "YES",
"NO");
data_frame(percent, gender_balance)
## # A tibble: 4 x 2
## percent gender_balance
## <dbl> <chr>
## 1 40 NO
## 2
      47 YES
## 3
      55 YES
## 4
      58 NO
my college dat <- college recent grads %>%
 select(major, men, women, median) %>%
 mutate(total = men + women,
    pct female = round((women / total)*100, 2),
    gender_balanced = ifelse(pct_female >= 45 & pct_female <=
55, yes="Yes", no="No"))
```

## Rename variables using rename():

Underscores are preferred over periods in variable names (but both work)

We can use rename() to change the name of gender.balanced to gender\_balanced.

rename([NEW VARIABLE NAME] = [OLD VARIABLE NAME])

```
my_college_dat <- my_college_dat %>%
 rename(median salary = median)
glimpse(my_college_dat)
```

head(n=3) only show the first 3 observation

<sup>\*</sup>will have new variables and new column for each variable

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### Missing values (coded as NA) in R:

The \$ notation is used to refer to a specific variable in a data frame college\_recent\_grads\$women is.na() to create a vector indicating TRUE where there is an NA and FALSE otherwise

is.na(college\_recent\_grads\$women)

# **Removing NAs from calculations:**

 Using na.rm=TRUE within the mean function removes the NA observations from the mean calculation

college\_recent\_grads %>% summarise(femgrad\_mean = mean(women, na.rm=TRUE), N=n())

Filtering just the observations that are **not NA** before doing the calculations

college\_recent\_grads %>%
filter(!is.na(women)) %>%
summarise(femgrad\_mean = mean(women), N=n())

summarise(temgrad\_mean = mean(women), N=n())