431 Class 04

Thomas E. Love

2019-09-05

Today's Agenda

Using R, RStudio and R Markdown and the 431 RStudio Cloud

Contact us at 431-help@case.edu

Our web site: https://github.com/THOMASELOVE/2019-431

A Worked Day 1 Survey Analysis

We have data on the site in a file called surveyday1_2019.csv. I built a R Markdown file, and then knitted it into an a PDF and an HTML then posted links to the Class 04 README.

- A "floating" table of contents
- Key verbs in the tidyverse for data wrangling
 - select, filter, count, arrange, mutate, group_by, summarize
- Building Histograms to describe a single quantitative variable
- Comparing a distribution of a quantity within groups
 - Faceted histogram
 - Comparison boxplot
- Obtaining numerical summaries
- Scatterplots with ggplot

All of this is also in the Course Notes.

Analyzing the Day 1 Survey

Load the R Packages we need

```
library(magrittr); library(tidyverse)
## always need tidyverse, can include other packages too
```

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Analyzing the Day 1 Survey

Load the Data

We will read in the .csv file of the data, and then pipe that result into the tbl_df function, which turns the data frame into a nicely organized *tibble*.

• Since we've stored the data file in the same directory as our R Project, we can read it in directly.

```
day1 <- read.csv("surveyday1_2019.csv") %>% tbl_df
```

Print your tibble by typing its name

day1

```
# A tibble: 315 x 21
  student sex glasses english statsofar ageguess
    <int> <fct> <fct> <fct>
                                    <int>
                                             <int>
   201901 <NA> y
                                        6
                                                42
 2 201902 <NA> v
                                                53
                        У
 3 201903 <NA> y
                                                45
                                        4
                        У
4 201904 <NA> y
                                                45
                        у
 5 201905 <NA> y
                                        6
                                                42
                        У
6 201906 <NA> v
                                                50
                        У
7 201907 <NA> y
                                        5
                                                56
8 201908 <NA> n
                                                50
                        n
   201909 <NA> n
                                        6
                                                52
                        V
10 201910 < NA > n
                                                42
                                        4
 ... with 305 more rows, and 15 more variables:
#
    smoke <int>, h.left <int>, h.right <int>,
```

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Use select to pick columns / variables from your tibble

```
day1 %>%
    select(favcolor, haircut)
# A tibble: 315 \times 2
  favcolor haircut
  <fct> <dbl>
 1 teal
            120
2 blue
            20
3 purple
          20
4 blue
 5 blue
             6.99
6 <NA>
             NA
 7 green
            25
8 red
             80
             16
 9 green
10 blue
              12.5
```

Use filter to pick rows / subjects from your tibble

```
day1 %>%
   filter(year == 2019)
 A tibble: 61 x 21
   student sex glasses english statsofar ageguess
     <int> <fct> <fct>
                        <fct>
                                    <int>
                                             <int>
   201901 <NA>
                                                42
                V
                         У
                                        6
 2 201902 <NA> y
                                                53
                         У
 3 201903 <NA> y
                                                45
4 201904 <NA>
                                                45
                V
                         у
 5 201905 <NA>
                                                42
                V
                         у
                                                50
   201906 <NA>
                V
                                                56
   201907 <NA>
                V
   201908 <NA>
                                                50
                        n
                                                52
   201909 <NA>
                                        6
                         V
10 201910 < NA > n
                                                42
                                        4
                         V
# ... with 51 more rows, and 15 more variables:
```

Use count to count the number of observations meeting a criterion

Or to provide a cross-classification:

```
day1 %>%
    count(favcolor == "blue", factor(english))
Warning: Factor `factor(english)` contains implicit NA,
consider using `forcats::fct_explicit_na`
# A tibble: 7 x 3
  `favcolor == "blue"` `factor(english)`
                                                n
  <lgl>
                        <fct>
                                            <int>
1 FALSE
                                               33
                        n
2 FALSE
                                              154
                        У
3 FALSE
                        < NA >
4 TRUE
                                               24
                        n
5 TRUE
                                               97
                        V
6 NA
                        n
                                                3
7 NA
                        V
```

```
day1 %>%
    mutate(guess_error = ageguess - lovetrueage) %>%
    select(ageguess, lovetrueage, guess_error) %>%
    summary()
```

ageg	guess	lovetrueage		guess_error	
Min.	:21.0	Min.	:47.50	Min.	:-31.500
1st Qu.	:45.0	1st Qu.	:48.50	1st Qu.	: -6.500
Median	:48.0	Median	:50.50	Median	: -2.500
Mean	:47.3	Mean	:50.13	Mean	: -2.837
3rd Qu.	:52.0	3rd Qu.	:51.50	3rd Qu.	: 0.500
Max.	:70.0	Max.	:52.50	Max.	: 20.500
NA's	:6			NA's	:6

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Get grouped summaries with group_by and summarize

```
day1 %>%
   group_by(year) %>%
   summarize(n = n(), average_guess = mean(ageguess),
        min_error = min(ageguess),
        max_error = max(ageguess),
        actual = mean(lovetrueage))
```

```
# A tibble: 6 x 6
           n average_guess min_error max_error actual
  <int> <int>
                     <dbl>
                               <dbl>
                                          <dbl>
                                                <dbl>
  2014
        42
                      NA
                                  NΑ
                                            NA 47.5
  2015
       49
                      47.1
                                  36
                                            57
                                                 48.5
3
  2016
        64
                      NA
                                  NΑ
                                            NA 49.5
  2017
        48
                      46.5
                                  29
                                            58 50.5
5
  2018
          51
                      NA
                                  NΑ
                                            NA
                                                 51.5
   2019
          61
                      NA
                                  NΑ
                                            NA
                                                 52.5
```

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Dealing with Missingness

We could filter our data to only include the subjects who provided a guess. . .

```
# A tibble: 6 x 6
            n average_guess min_error max_error actual
                        <dbl>
  <int> <int>
                                  <int>
                                             <int>
                                                     <dbl>
   2014
           41
                        47.3
                                      38
                                                 58
                                                      47.5
   2015 49
                        47.1
                                      36
                                                 57 48.5
3
   2016
        61
                        46.0
                                      24
                                                 70
                                                      49.5
                         46.5
                                      29
   2017
            48
                                                 58
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```

Looking at Errors, instead

What if, instead, we wanted to look at the errors made, by subtracting off my true age from everyone's guess?

```
# A tibble: 6 \times 5
  year n average_error min_error max_error
                <dbl>
                         <dbl>
 <int> <int>
                                  <dbl>
 2014 41
                -0.159 -9.5 10.5
2 2015 49
               -1.38 -12.5 8.5
3
  2016 61
                -3.53 -25.5 20.5
                 -3.96
                         -21.5
                                   7.5
  2017
        48
```

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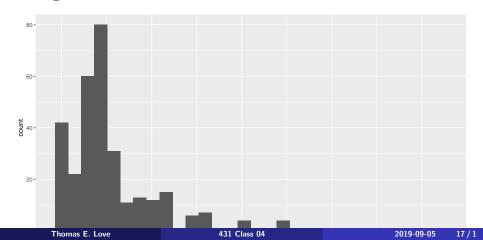
Histogram of Haircut Prices

```
ggplot(data = day1, aes(x = haircut)) +
    geom_histogram()
```

Histogram of Haircut Prices (Result)

`stat_bin()` using `bins = 30`. Pick better value
with `binwidth`.

Warning: Removed 4 rows containing non-finite values (stat_bin).

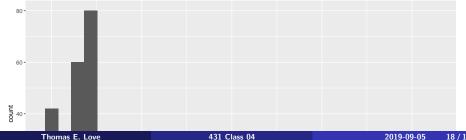


Improvements

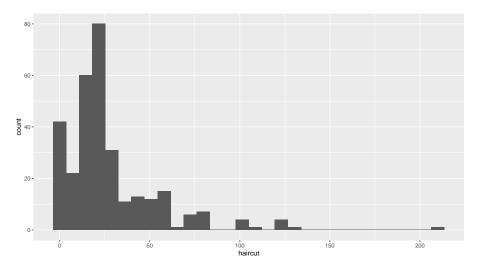
We'll filter the rows of the day1 tibble to include only those subjects who gave us a haircut price.

```
day1 %>%
    filter(complete.cases(haircut)) %>%
    ggplot(data = ., aes(x = haircut)) +
    geom_histogram()
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



`stat_bin()` using `bins = 30`. Pick better value
with `binwidth`.



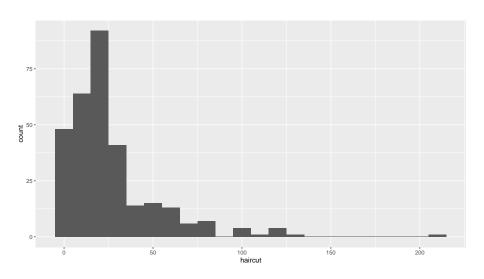
Improvements

We'll specify that R should create bins of width \$10 (rather than the default, which creates 30 bins) for the haircut prices to fall in.

```
day1 %>%
    filter(complete.cases(haircut)) %>%
    ggplot(data = ., aes(x = haircut)) +
    geom_histogram(binwidth = 10)
```

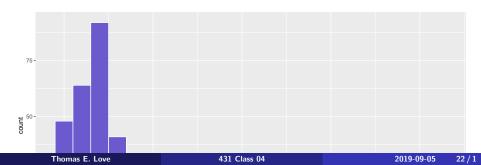


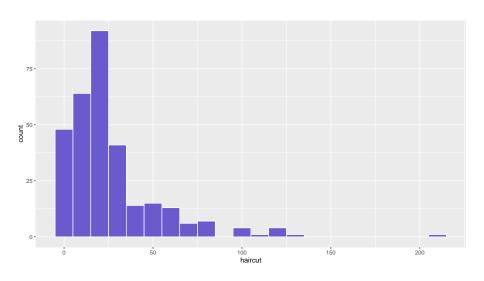
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Improvements

We'll set the fill to be a better color - a nice resource for this is to google Colors in R. I'll pick "slateblue". We'll also color the outlines of the bars "white".

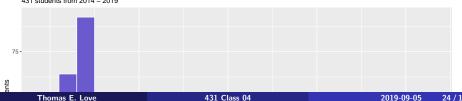


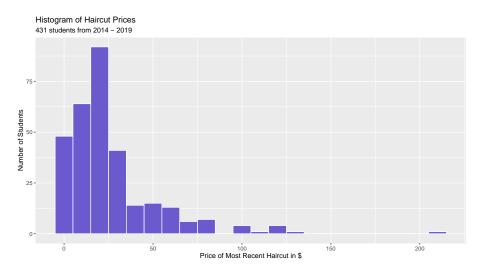


Improvements

We'll build a main title, subtitle and proper axis titles.

Histogram of Haircut Prices

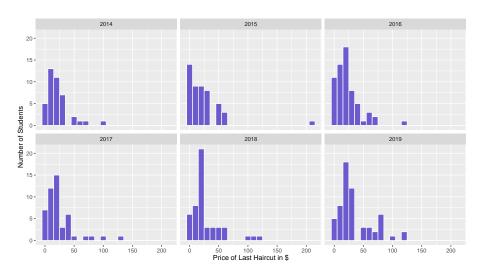




Separate histograms for each year with faceting?

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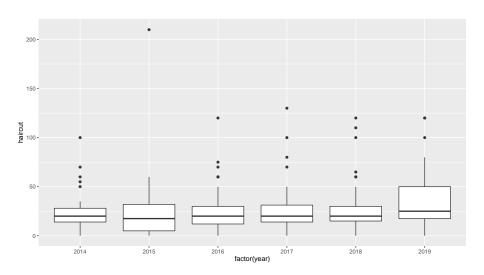
Resulting (using facet_wrap) plot



Building a Comparison Boxplot

We could use a comparison boxplot. A trick here is to specify year as a factor. . .

Comparison Boxplot



```
day1 %>%
    select(haircut) %>%
    summary()
    haircut
 Min. : 0.00
 1st Qu.: 14.00
 Median : 20.00
 Mean : 27.32
 3rd Qu.: 32.00
 Max. :210.00
 NA's :4
which can also be done with
summary(day1$haircut)
```

The mosaic package has a useful favstats function...

```
mosaic::favstats(day1$haircut)
```

But to get this in a pipeline, you'd need the %\$% operator from the magrittr package...

```
day1 %$%
  mosaic::favstats(haircut)
```

```
Registered S3 method overwritten by 'mosaic':

method from
fortify.SpatialPolygonsDataFrame ggplot2

min Q1 median Q3 max mean sd n missing
0 14 20 32 210 27.3199 26.35565 311 4
```

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The psych package has a useful describe function...

```
day1 %$%
    psych::describe(haircut)
```

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The Hmisc package also has a useful describe function...

```
day1 %$%
  Hmisc::describe(haircut)
```

haircut

Gmd	Mean	${\tt Info}$	distinct	missing	n
25.36	27.32	0.992	50	4	311
.90	.75	.50	.25	.10	.05
60	32	20	14	0	0
					.95
					80

```
lowest: 0.0 1.0 3.0 3.5 5.0 highest: 100.0 110.0 120.0 130.0 210.0
```

Numerical Summary by Year?

Numerical Summary by Year (Result)

```
A tibble: 6 x 5
                        sd median
   year
            n
               mean
  <int> <int> <dbl> <dbl>
                            <dbl>
   2014
           41
               23.7
                      19.8
                             20
2
   2015
           49
               24.7 32.8
                          17.5
3
           63 23.8 21.5
   2016
                             20
4
   2017
           48
               25.9
                    25.3
                             20
5
   2018
           50
               28.3
                    26.1
                             20
6
   2019
           60
               35.8
                             25
                     29.1
```

What is the relationship between 431 students' pulse rate and hours of sleep the prior night?

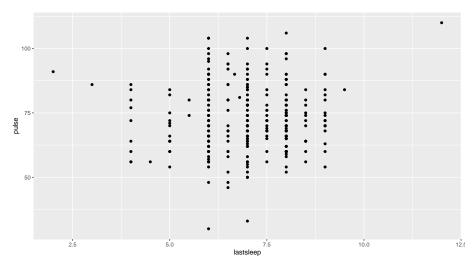
Here, we're looking at two quantitative variables. A **scatterplot** is usually the best choice.

```
ggplot(data = day1, aes(x = lastsleep, y = pulse)) +
    geom_point()
```

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Scatterplot (Result)

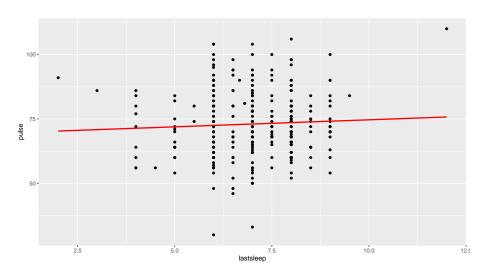
Warning: Removed 3 rows containing missing values (geom_point).



Improving the Scatterplot

Let's filter to include only those cases with known pulse and known lastsleep, and also add a line from a linear regression model to predict pulse rate on the basis of hours of sleep the prior night.

```
day1 %>%
  filter(complete.cases(pulse, lastsleep)) %>%
  ggplot(data = ., aes(x = lastsleep, y = pulse)) +
  geom_point() +
  geom_smooth(method = "lm", se = FALSE, col = "red")
```



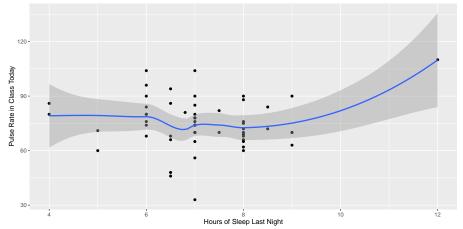
Smoothing the 2019 data

Let's look at the 2019 data only, and fit a curved (loess) smooth to predict pulse rate on the basis of hours of sleep the prior night. We'll also add a title and subtitle and retitle the axes

```
day1 %>%
  filter(year == "2019") %>%
  filter(complete.cases(pulse, lastsleep)) %>%
  ggplot(data = ., aes(x = lastsleep, y = pulse)) +
  geom_point() +
  geom_smooth(method = "loess") +
  labs(title = "Pulse Rate as a Function of Hours of Sleep I
      subtitle = "with fitted loess smooth, students in the
      x = "Hours of Sleep Last Night",
      y = "Pulse Rate in Class Today")
```

The Results

Pulse Rate as a Function of Hours of Sleep Last Night with fitted loess smooth, students in the 2019 class



A Linear Model?

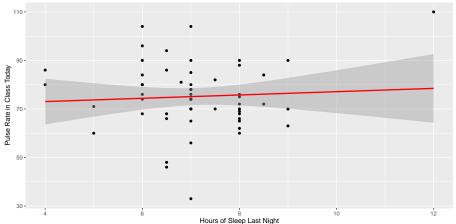
We could instead restrict ourselves to a linear model for the 2019 group.

```
day1 %>%
    filter(year == "2019") %>%
    filter(complete.cases(pulse, lastsleep)) %>%
    ggplot(data = ., aes(x = lastsleep, y = pulse)) +
    geom point() +
    geom smooth(method = "lm", col = "red") +
    labs(title = "Pulse Rate as a Function of Hours of Sleep |
         subtitle = "with fitted linear model, students in the
         x = "Hours of Sleep Last Night",
         y = "Pulse Rate in Class Today")
```

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Linear Fit (Results)

Pulse Rate as a Function of Hours of Sleep Last Night with fitted linear model, students in the 2019 class



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Correlation?

The correlation of lastsleep and pulse is likely to be of some interest. Note the use of both the %>% and %\$% pipes in this case.

```
day1 %>%
  filter(year == "2019") %>%
  filter(complete.cases(pulse, lastsleep)) %$%
  cor(pulse, lastsleep)
```

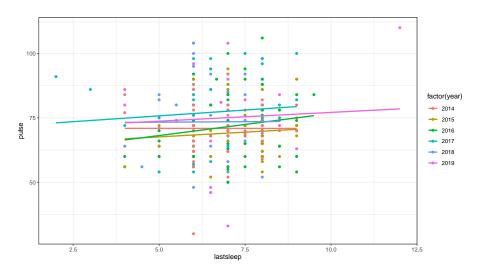
[1] 0.06356228

Does the linear model change much by year?

Here's the plot, color coding the models by year (note the use of the group as well as the color aesthetic here), and also incorporating the black-and-white theme, rather than the default.

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Does the linear model change much by year?



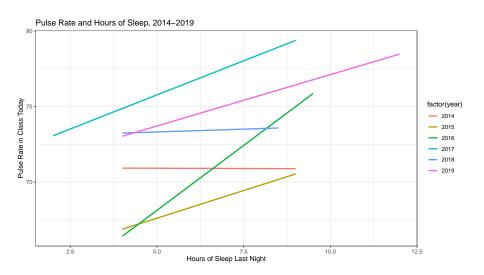
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Plot of the models only

Here's the same plot of the models alone, and not showing the data (commenting out the line of code that draws the points.) We'll also improve the labeling.

```
day1 %>%
    filter(complete.cases(pulse, lastsleep)) %>%
    ggplot(., aes(x = lastsleep, y = pulse,
                                color = factor(year),
                                group = factor(year))) +
  geom\ point() +
 #
    geom smooth(method = "lm", se = FALSE) +
    labs(title = "Pulse Rate and Hours of Sleep, 2014-2019",
         x = "Hours of Sleep Last Night",
         y = "Pulse Rate in Class Today") +
      theme bw()
```

Plot of the models only

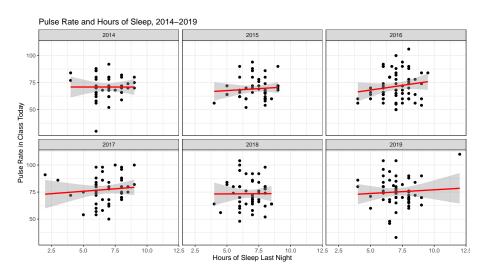


Faceting a Scatterplot

Here's the same basic information, but faceted by year.

```
day1 %>%
    filter(complete.cases(pulse, lastsleep)) %>%
    ggplot(data = ., aes(x = lastsleep, y = pulse,
                                group = factor(year))) +
    geom point() +
    geom smooth(method = "lm", color = "red") +
    facet wrap(~ year) +
    labs(title = "Pulse Rate and Hours of Sleep, 2014-2019",
         x = "Hours of Sleep Last Night",
         y = "Pulse Rate in Class Today") +
    theme bw()
```

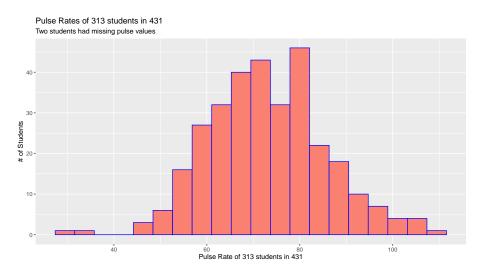
Faceting a Scatterplot



Analyzing the Survey Data - A little challenge

Can you reproduce the following...

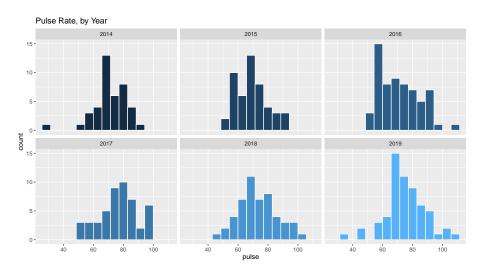
A. That fill color is called *salmon*, I used 20 bins.



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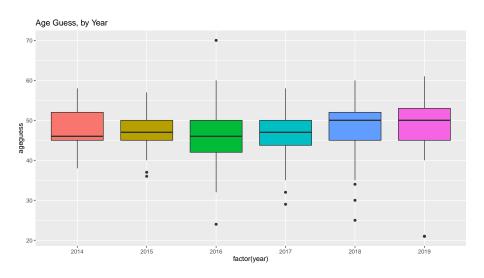
```
day1 %>% filter(complete.cases(pulse)) %>%
  ggplot(data = ., aes(x = pulse)) +
  geom_histogram(bins = 20, col = "blue", fill = "salmon") +
  labs(x = "Pulse Rate of 313 students in 431",
        y = "# of Students",
        title = "Pulse Rates of 313 students in 431",
        subtitle = "Two students had missing pulse values")
```

B. Histograms of Pulse Rates, Faceted by Year



```
day1 %>% filter(complete.cases(pulse)) %>%
  ggplot(data = ., aes(x = pulse, fill = year)) +
  geom_histogram(bins = 15, col = "white") +
  facet_wrap(~ year) +
  guides(fill = FALSE) +
  labs(title = "Pulse Rate, by Year")
```

C. Boxplots of Age Guesses, by Year



Summary Table of Age Guesses, by Year

```
A tibble: 6 x 5
                   sd median
              mean
  year
           n
 <int> <int> <dbl> <dbl>
                        <dbl>
              47.3 5.21
  2014
          41
                            46
2
  2015
       49 47.1 4.62
                            47
3
  2016
       61 46.0 7.00
                            46
4
  2017
       48 46.5 6.15
                            47
5
  2018
          50
              48.2 6.47
                            50
6
  2019
          60
              48.6 7.09
                            50
```

Code for Summary Table