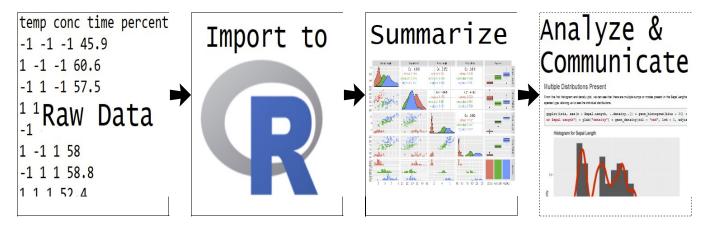
NC STATE UNIVERSITY

Summarizing Data: Descriptive Statistics

What is this course about?

Basic use of R for reading, manipulating, and plotting data!



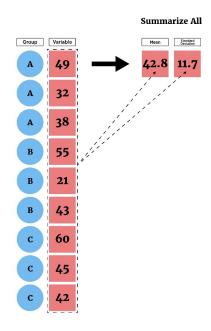
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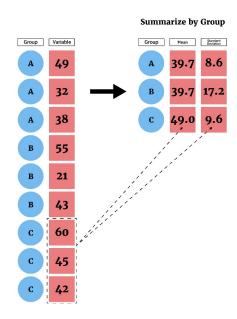
- · read and write basic R programs
- · import well formatted data into R
- · do basic data manipulation in R
- · produce common numerical and graphical summaries in R
- · describe a use case of an analysis done in R

· Understand types of data and their distributions

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- Numerical summaries (across subgroups)



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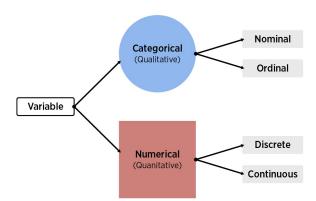


- · Understand types of data and their distributions
- · Numerical summaries (across subgroups)
 - Contingency Tables
 - Mean/Median
 - Standard Deviation/Variance/IQR
 - Quantiles/Percentiles

- · Understand types of data and their distributions
- Numerical summaries (across subgroups)
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- Graphical summaries (across subgroups)
 - Bar plots
 - Histograms
 - Box plots
 - Scatter plots

Understanding Data

- · How to summarize data?
- · Depends on data type:
 - Categorical (Qualitative) variable entries are a label or attribute
 - Numeric (Quantitative) variable entries are a numerical value where math can be performed



Understanding Data

Common goal: Describe the distribution of the variable

- · Distribution = pattern and frequency with which you observe a variable
- · Categorical variable describe relative frequency (or count) in each category
- · Numeric variable describe the shape, center, and spread

Contingency tables

Categorical variable - entries are a label or attribute

- · Tables (contingency tables) via table
 - Show frequency/proportion of categories

Contingency tables

Categorical variable - entries are a label or attribute

- · Tables (contingency tables) via table
 - Show frequency/proportion of categories
- Consider data on titanic passengers in titanic.csv

```
## # A tibble: 1,310 x 14
    pclass survived name sex
                                    age sibsp parch ticket fare cabin embarked
      <dbl>
               <dbl> <chr> <dbl> <dbl> <dbl> <dbl> <chr> <dbl> <chr> <</pre>
##
                                            0
                                                            211. B5
## 1
                   1 Alle~ fema~ 29
                                                  0 24160
                                                  2 113781 152. C22 ~ S
## 2
                   1 Alli~ male
                                0.917
                                            1
                                                  2 113781 152. C22 ~ S
## 3
                   O Alli~ fema~ 2
                                            1
                   0 Alli~ male 30
                                                  2 113781 152. C22 ~ S
## 4
                                            1
## 5
                   0 Alli~ fema~ 25
                                            1
                                                  2 113781 152. C22 ~ S
## # ... with 1,305 more rows, and 3 more variables: boat <chr>, body <dbl>,
      home.dest <chr>
## #
```

Contingency tables

- · Create **one-way contingency tables** for each of three categorical variables:
 - embarked (where journey started)
 - survived (survive or not)
 - sex (Male or Female)

```
table(titanicData$embarked)
                                                  table(titanicData$survived)
##
                                                  ##
                                                  ##
                                                       0
    С
        Q
                                                           1
## 270 123 914
                                                  ## 809 500
                                                  table(titanicData$sex)
                                                  ##
                                                  ## female male
                                                        466
                                                               843
```

Two-way contingency tables

· Create two-way contingency tables for pairs of categorical variables

table(titanicData\$survived, table(titanicData\$survived, titanicData\$sex) titanicData\$embarked) ## ## ## female male ## С Q S 127 682 ## ## 0 120 79 610 339 161 1 150 44 304 ## ## table(titanicData\$sex, titanicData\$embarked) ## ## С Q S ## female 113 60 291 male 157 63 623 ##

Three-way contingency tables

· Create a **three-way contingency table** for three categorical variables

table(titanicData\$sex, titanicData\$embarked, titanicData\$survived)

Three-way contingency tables

- Create a three-way contingency table for three categorical variables (order matters for output!)
- Example of an array! 3 dimensions [, ,]

```
tab <- table(titanicData$sex, titanicData$embarked, titanicData$survived)
str(tab)

## 'table' int [1:2, 1:3, 1:2] 11 109 23 56 93 517 102 48 37 7 ...
## - attr(*, "dimnames")=List of 3
## ..$ : chr [1:2] "female" "male"
## ..$ : chr [1:3] "C" "Q" "S"
## ..$ : chr [1:2] "0" "1"</pre>
```

· Can obtain **conditional** bivariate info!

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· Can obtain **conditional** bivariate info!

· Can obtain **conditional** univariate info too!

```
## 'table' int [1:2, 1:3, 1:2] 11 109 23 56 93 517 102 48 37 7 ...
## - attr(*, "dimnames")=List of 3
## ..$ : chr [1:2] "female" "male"
## ..$ : chr [1:3] "C" "Q" "S"
## ..$ : chr [1:2] "0" "1"

#Survived status for males that embarked at "Q"
tab[2, 2, ]

## 0 1
## 56 7
```

Numeric variable - entries are a numerical value where math can be performed **Single variable:**

- · Shape: Histogram or Density plot
- · Measures of center: Mean, Median
- · Measures of spread: Variance, Standard Deviation, Quartiles, IQR

Numeric variable - entries are a numerical value where math can be performed

Single variable:

- · Shape: Histogram or Density plot
- · Measures of center: Mean, Median
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Two Variables:

- · Shape: Scatter plot
- · Measures of linear relationship: Covariance, Correlation

- · Look at carbon dioxide (CO2) uptake data set
 - Response recorded: uptake CO2 uptake rates in grass plants
 - Environment manipulated: Treatment chilled/nonchilled
 - Ambient CO2 specified and measured: conc

```
CO2 <- tbl df(CO2)
CO2
## # A tibble: 84 x 5
    Plant Type Treatment conc uptake
    <ord> <fct> <fct>
                       <dbl> <dbl>
## 1 Qn1
         Quebec nonchilled 95
                                 16
                            175
## 2 Qn1 Quebec nonchilled
                                 30.4
## 3 Qn1 Quebec nonchilled 250
                                 34.8
## 4 Qn1 Quebec nonchilled 350
                                 37.2
## 5 Qn1 Quebec nonchilled
                                 35.3
                            500
## # ... with 79 more rows
```

Measures of center

Mean & Median

```
mean(CO2$uptake)

## [1] 27.2131

#note you can easily get a trimmed mean
mean(CO2$uptake, trim = 0.05) #5% trimmed mean

## [1] 27.25263

median(CO2$uptake)

## [1] 28.3
```

Measures of spread

Variance, Standard Deviation, Quartiles, & IQR

```
#quartiles and mean
summary(CO2$uptake)
##
     Min. 1st Qu. Median Mean 3rd Qu.
                                             Max.
     7.70 17.90
                   28.30 27.21
                                    37.12
                                            45.50
##
var(CO2$uptake)
                                       IQR(CO2$uptake)
## [1] 116.9515
                                       ## [1] 19.225
                                       quantile(CO2$uptake, probs = c(0.1, 0.2))
sd(CO2$uptake)
## [1] 10.81441
                                       ##
                                            10%
                                                  20%
                                       ## 12.36 15.64
```

Measures of linear relationship

Covariance & Correlation

```
cov(CO2$conc, CO2$uptake)
## [1] 1552.687

cor(CO2$conc, CO2$uptake)
## [1] 0.4851774
```

Usually want summaries for different subgroups of data

• Ex: Get similar uptake summaries for each **Treatment**

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Idea:

- Use group_by to create subgroups associated with the data frame
- Use summarize to create basic summaries for each subgroup

• Ex: Get similar uptake summaries for each **Treatment**

• Ex: Get similar uptake summaries for each **Treatment** and **Concentration**

```
CO2 %>% group by (Treatment, conc) %>%
       summarise(avg = mean(uptake), med = median(uptake), var = var(uptake))
## # A tibble: 14 x 5
## # Groups: Treatment [2]
     Treatment conc
##
                     avg
                         med
                                var
     ##
                95 13.3 12.8 5.75
  1 nonchilled
  2 nonchilled 175 25.1 24.6 32.6
  3 nonchilled 250 32.5 32.7 35.1
## 4 nonchilled
                350 35.1 34.5 37.4
## 5 nonchilled
                500 35.1 33.8 31.9
## 6 nonchilled
                675 36.0 35.8 40.2
## 7 nonchilled 1000 37.4 37.6 49.8
## 8 chilled
                 95 11.2 10.6 8.18
  9 chilled
                175 19.4 19.5 34.7
## 10 chilled
                250 25.3 24.2 112.
## 11 chilled
                350 26.2 26.4 117.
## 12 chilled
                500 26.6 26
                              131.
## 13 chilled
                675 27.9 28.8 120.
## 14 chilled
               1000 29.8 30.3 154.
```

dplyr has variations on summarise that can be used:

- summarise_all() Apply functions to every column
- summarise_at() Apply functions to specific columns
- summarise_if() Apply functions to all columns of one type

- Ex: Get similar uptake summaries for each **Treatment**
- Built-in aggregate () function more general

- Ex: Get similar uptake summaries for each **Treatment**
- Built-in aggregate() function more general
- \cdot Basic use gives response (x) and a list of variables to group by

```
aggregate(x = CO2$uptake, by = list(CO2$Treatment), FUN = summary)

## Group.1 x.Min. x.1st Qu. x.Median x.Mean x.3rd Qu. x.Max.

## 1 nonchilled 10.60000 26.47500 31.30000 30.64286 38.70000 45.50000

## 2 chilled 7.70000 14.52500 19.70000 23.78333 34.90000 42.40000
```

- Ex: Get similar uptake summaries for each **Treatment**
- Built-in aggregate () function more general

aggregate (uptake ~ Treatment, data = CO2, FUN = summary)

Commonly used with formula notation!

7.70000

```
## Treatment uptake.Min. uptake.1st Qu. uptake.Median uptake.Mean
## 1 nonchilled 10.60000 26.47500 31.30000 30.64286
```

14.52500

19.70000

23.78333

uptake.3rd Qu. uptake.Max. ## 1 38.70000 45.50000

chilled

2

2 34.90000 42.40000

- Ex: Get similar uptake summaries for each **Treatment**
- Built-in aggregate() function more general
- Commonly used with formula notation!

```
aggregate (uptake \sim Treatment, data = CO2, FUN = summary) uptake \sim Treatment - formula notation in R
```

· Idea: uptake (LHS) modeled by Treatment levels (RHS)

- Ex: Get similar uptake summaries for each **Treatment** and **Concentration**
- Built-in aggregate() function more general
- Commonly used with formula notation!

```
aggregate (uptake ~ Treatment + conc, data = CO2, FUN = summary)
```

uptake ~ Treatment + conc model uptake by levels of Treatment and conc

 $\cdot\;$ Ex: Get similar uptake summaries for each Treatment and Concentration

aggregate (uptake ~ Treatment + conc, data = CO2, FUN = summary)

##				-	uptake.1st Qu.	-	-
	1	nonchilled	95	10.60000	11.47500	12.80000	13.28333
##		chilled	95	7.70000	9.60000	10.55000	11.23333
##	3	nonchilled	175	19.20000	20.05000	24.65000	25.11667
##	4	chilled	175	11.40000	15.67500	19.50000	19.45000
##	5	nonchilled	250	25.80000	27.30000	32.70000	32.46667
##	6	chilled	250	12.30000	17.95000	24.20000	25.28333
##	7	nonchilled	350	27.90000	30.45000	34.50000	35.13333
##	8	chilled	350	13.00000	18.15000	26.45000	26.20000
##	9	nonchilled	500	28.50000	31.27500	33.85000	35.10000
##	10	chilled	500	12.50000	18.30000	26.00000	26.65000
##	11	nonchilled	675	28.10000	31.42500	35.80000	36.01667
##	12	chilled	675	13.70000	19.72500	28.80000	27.88333
##	13	nonchilled	1000	27.80000	32.50000	37.60000	37.38333
##	14	chilled	1000	14.40000	20.40000	30.30000	29.78333
##		uptake.3rd Qu. uptake.Max.					
##	1	15.40	0000	16.20000			
##	2	13.30	0000	15.10000			
##	3	29.62	2500	32.40000			
##	4	23.32500		27.30000			
	5	36.52500		40.30000			
	6	33.82500		38.10000			
##	7	40.65000		42.10000			
##	8	34.45000		38.80000			
##		39.27500		42.90000			
	10	37.07500		38.90000			
	11	40.85000		43.90000			
##	12	36.97500		39.60000			
	13	43.15		45.50000			
	14	40.72		42.40000			
11 11	т 1	10.72		12.10000			

Recap/Next Up!

- · Understand types of data and their distributions
- · Numerical summaries
 - Contingency Tables: table
 - Mean/Median: mean, median
 - Standard Deviation/Variance/IQR: sd, var, IQR
 - Quantiles/Percentiles: quantile
- Across subgroups with dplyr::group_by and dplyr::summarize or aggregate
- Graphical summaries (across subgroups)