# Clinical statistics for non-statisticians: Day one

**Steve Simon** 

# One warning

- Lots of real world analogies, but
  - May be too specific to U.S.A.
  - Please ask about anything obscure

## Start with a bad joke

Put your reaction ("Ha ha", "Groan", etc.) in the chat box. Two statistics are sitting in a bar. One turns to the other and asks, "So, how do you like married life?" The other statistic responds ...

#### Introduction

- Tell us one interesting number about yourself
- Examples
  - 8: I have traveled to eight countries outside the United States
    - (Canada, Italy, China, France, Russia, England, Holland, and Iceland)
  - 29: I did not learn how to drive until I was 29 years old
  - 1802: My highest chess rating was 1802, but I am not that good any more.

## **Your turn**

## A bit more about myself

- PhD in Statistics in 1982 from the University of Iowa
- Currently full professor
- Part-time statistical consultant
- Funded on 18 research grants
- Over 100 peer-reviewed publications
- Website with over 2,000 pages
- Many invitations to talk at conferences

### Outline of the three day course

- Day one: Numerical summaries and data visualization
- Day two: Hypothesis testing and sampling
- Day three: Statistical tests to compare treatment to a control and regression models

My goal: help you to become a better consumer of statistics

## Day one topics

- Numerical summaries
  - When should you present the mean versus the median
  - When should you present the range versus standard deviation
  - How should you display percentages
  - Why should you round liberally

## Day one topics (continued)

- Data visualization
  - How should you display continuous data
  - Why is the normal bell-shaped curve important
  - How should you display categorical data
  - How do you illustrate trends and patterns
  - What are some common mistakes in the choice of colors

## Quiz question 1

```
No
                         Yes
                                       Total
1
           154 (33%)
                       308 (67%)
                                   462 (100%)
2 Female
3 Male
                       142 (17%)
           709 (83%)
                                   851 (100%)
           863 (66%)
                       450 (34%)
4 Total
                                  1313 (100%)
```

This data table shows counts and ...

- 1. cell percents
- 2. column percents
- 3. row percents
- 4. I do not know the answer

# **Quiz question 2**

The median might be preferred to the mean if

- 1. a single extreme value distorts the mean
- 2. the data follows a bell shaped curve
- 3. there is very little variation in the data
- 4. you have a biased sample
- 5. I do not know the answer

## **Quiz question 3**

The problem with error bars is that they

- 1. fail to show if the data is skewed
- 2. have several competing definitions
- 3. use only two numbers to characterize your data
- 4. all of the above are correct
- 5. none of the above are correct
- 6. I do not know the answer

# **Counting and proportions**

- Counts are the most common statistic
  - Counts are error prone
  - Counts require a solid operational definition

#### Student exercise

Count the number of occurrences of the letter "e".

A quality control program is easiest to implement from the top down. Make sure that you understand the the commitment of time and money that is involved. Every workplace is different, but think about allocating 10% of your time and 10% of the time of all your employees to quality control.

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Figure 1: Image of a haemocytometer

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Figure 2: Titanic data: counts of survival by gender

Figure 3: Titanic data with column percentages

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Figure 4: Titanic data with row percentages

# Percentages divided by grand total

Figure 5: Titanic data with cell percentages

## My recommendations

- Treatment or exposure as rows
- Outcome as columns
- Usually report row percentages
  - Female survival rate: 67%
  - Male survival rate: 17%
- But sometimes column percentages
  - Survivors: 68% female, 32% male

#### Some rationale for these choices

#### My way

```
Survived
No
Yes
Sex Female 33%
(154) 67% (308)
Male 83%
(863) 17% (142)
```

#### Not my way

```
Sex
Female
Male
Survived No 33%
(154) 83% (863)
Yes 67%
(308) 17% (142)
```

#### **Break**

- What have you just learned?
  - Displaying percentages
- What is coming next?
  - Practice exercise
  - Calculation of the mean and median

## On your own

Calculate row and column percentages for the following tables. Interpret your results.

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Figure 6: Titanic passenger class counts

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Figure 7: Titanic child

counts

.

Figure 8: Cartoon image of Professor Mean

.

Figure 9: Road with a median strip

#### Calculation of the mean and median

- Mean
  - Add up all the values, divide by the sample size
- Median
  - Sort the data
    - Select the middle value if n is odd
    - go halfway between the two middle values if n is even

#### Formal mathematical definitions

- Mean
  - $lacksquare ar{X} = rac{1}{n} \Sigma X_i$
- Median
  - lacksquare Sorted values  $X_{[1]}, X_{[2]}, \ldots, X_{[n]}$ 
    - $\circ \ X_{[(n+1)/2]}$  if n is odd,
    - $\circ \ (X_{[n/2]} + X_{[n/2+1]})/2$  if n is even

# **Bacteria before and after A/C upgrade**

Room	Before	After
121	11.8	10.1
125	7.1	3.8
163	8.2	7.2
218	10.1	10.5
233	10.8	8.3
264	14	12
324	14.6	12.1
325	14	13.7

## **Before remediation mean**

```
11.8 + 7.1 + 8.2 + 10.1 + 10.8 + 14 + 14.6 + 14 = 90.6

90.6 / 8 = 11.325

Round to 11.3
```

## After remediation mean

$$10.1 + 3.8 + 7.2 + 10.5 + 8.3 + 12 + 12.1 + 13.7$$
  
= 77.7

$$77.7 / 8 = 9.7125$$

Round to 9.7

# Before remediation median (1/4)

```
121 11.8
```

125 7.1

163 8.2

218 10.1

233 10.8

264 14.0

324 14.6

325 14.0

# Before remediation median (2/4)

```
125 7.1
```

163 8.2

218 10.1

233 10.8

121 11.8

264 14.0

325 14.0

324 14.6

# Before remediation median (3/4)

```
125 7.1
```

163 8.2

218 10.1

233 10.8 10.8

121 11.8 11.8

264 14.0

325 14.0

324 14.6

# Before remediation median (4/4)

```
125
     7.1
163 8.2
    10.1
218
    10.8
233
          10.8
                 (10.8 + 11.8) / 2 = 11.3
121
    11.8
          11.8
264
    14.0
325
    14.0
324 14.6
```

# After remediation median (1/4)

```
121 10.1
```

125 3.8

163 7.2

218 10.5

233 8.3

264 12.0

324 12.1

325 13.7

# After remediation median (2/4)

- 125 3.8
- 163 7.2
- 233 8.3
- 121 10.1
- 218 10.5
- 264 12.0
- 324 12.1
- 325 13.7

## After remediation median (3/4)

```
125 3.8
```

163 7.2

233 8.3

121 10.1 10.1

218 10.5 10.5

264 12.0

324 12.1

## After remediation median (4/4)

```
125 3.8

163 7.2

233 8.3

121 10.1 10.1

(10.1 + 10.5) / 2 = 10.3

218 10.5 10.5

264 12.0

324 12.1

325 13.7
```

#### **Break**

- What have you just learned?
  - Calculation of the mean and median
- What is coming next?
  - Criticisms of the mean and median

## Criticisms of the mean and median

- Are you combining apples and onions?
- Are you ignoring minorities?

#### Use of the mean for ordinal data

- Stevens scales of measurement (controversial!)
  - Nominal
  - Ordinal
  - Interval
  - Ratio
- Addition/subtraction not allowed for ordinal data
  - Mean of ordinal data is meaningless

## An example of ordinal data.

- "Do you agree or disagree with the following statements"
  - "I believe that knowledge of Statistics is important for my job."
    - ∘ 1 = Strongly disagree,
    - ∘ 2 = Disagree
    - ∘ 3 = Neutral
    - 4 = Agree
    - ∘ 5 = Strongly agree

# Another example of ordinal data, course grades

- A = 4
- B = 3
- C = 2
- D = 1
- F = 0

Figure 10: Excerpt from Gould 1985 publication

#### Choosing between the mean and median

- Often, either is fine
- When do you use the mean?
  - When totals are important
  - "In 2020, the average expenditure by the Italian National Health Service (Servizio Sanitario Nazionale, SSN) per patient affected by at least one chronic disease was approximately 696 euros."
- When do you use the median
  - When outliers/skewness might distort your conclusions

Figure 11: Chen et al 2019

#### Chen 2019, PMID: 31806195 (continued)

Background: The prices of newly approved cancer drugs have risen over the past decades. A key policy question is whether the clinical gains offered by these drugs in treating specific cancer indications justify the price increases.

#### Chen 2019, PMID: 31806195 (continued)

Results: We found that between 1995 and 2012, price increases outstripped median survival gains, a finding consistent with previous literature. Nevertheless, price per mean life-year gained increased at a considerably slower rate, suggesting that new drugs have been more effective in achieving longer-term survival.

Between 2013 and 2017, price increases reflected equally large gains in median and mean survival, resulting in a flat profile for benefit-adjusted launch prices in recent years.

## **Break**

- What have you just learned?
  - Criticisms of the mean and median
- What is coming next?
  - Computing percentiles

Figure 12: Illustration of the 75th percentile

## **Computing percentiles**

- Many formulas
  - Differences are not worth fighting over
- My preference (pth quantile)
  - Sort the data
  - Calculate p\*(n+1)
  - Is it a whole number?
    - Yes: Select that value, otherwise
    - No: Go halfway between
    - ∘ Special cases: p(n+1) < 1 or > n

## Some examples of percentile calculations

- Example for n=39
  - For 5th percentile, p(n+1)=2 -> 2nd smallest value
  - For 4th percentile, p(n+1)=1.6 -> halfway between two smallest values
  - For 2nd percentile, p(n+1)=0.8 -> smallest value

## Some terminology

- Percentile: goes from 0% to 100%
- Quantile: goes from 0.0 to 1.0
  - 90th percentile = 0.9 quantile
- 25th, 50th, and 75th percentiles: quartiles
  - ullet 25th percentile:  $Q_1,\ X_{0.25}$  or lower quartile
  - ullet Median/50th percentiles:  $Q_2$  or  $X_{0.5}$
  - 75th percentile:  $Q_3,\ X_{0.75}$  or upper quartile

# Before remediation upper quartile (1/4)

```
121 11.8
```

125 7.1

163 8.2

218 10.1

233 10.8

264 14.0

324 14.6

325 14.0

# Before remediation upper quartile (2/4)

```
125 7.1
```

163 8.2

218 10.1

233 10.8

121 11.8

264 14.0

325 14.0

324 14.6

## Before remediation upper quartile (3/4)

```
125 7.1
```

163 8.2

218 10.1

233 10.8

121 11.8

264 14.0 14

325 14.0 14

324 14.6

## Before remediation upper quartile (4/4)

```
125
     7.1
163 8.2
    10.1
218
    10.8
233
    11.8
121
264
    14.0
         14
                 (14 + 14) / 2 = 14
325
    14.0
          14
324 14.6
```

# After remediation upper quartile (1/4)

```
121 10.1
```

125 3.8

163 7.2

218 10.5

233 8.3

264 12.0

324 12.1

# After remediation upper quartile (2/4)

```
125 3.8
```

163 7.2

233 8.3

121 10.1

218 10.5

264 12.0

324 12.1

## After remediation upper quartile (3/4)

```
125 3.8
```

163 7.2

233 8.3

121 10.1

218 10.5

264 12.0 12

324 12.1 12.1

## After remediation upper quartile (4/4)

## When you should use percentiles

- Characterize variation
  - Middle 50% of the data
- Exposure issues
  - Not enough to control median exposure level
- Quantify extremes
  - What does "upper class" mean?
- Quality control
  - Almost all products must meet a minimum standard

## **Break**

- What have you just learned?
  - Computing percentiles
- What is coming next?
  - Computing the standard deviation

## **Standard deviation**

$$S = \sqrt{rac{1}{n-1} \Sigma (X_i - ar{X})^2}$$
 At least one alternative formulas.

## Why is variation important

- Variation = Noise
  - Too much noise can hide signals
- Variation = Heterogeneity
  - Too little heterogeneity, hard to generalize
  - Too much heterogeneity, mixing apples and oranges
- Variation = Unpredictability
  - Too much unpredictability, hard to prepare for the future
- Variation = Risk
  - Too much risk can create a financial burden

## Should you try to minimize variation?

- Yes, for early studies
  - Easier to detect signals
  - Proof of concept trials
- No, for later studies
  - Easier to generalize results
  - Pragmatic trials

#### The bell shaped curve

- Does your variation follow a bell shaped curve?
- Synonyms: normality, normal distribution
  - Values in the middle are most common
  - Frequencies taper off away from the center
  - Symmetry on either side
- A bell shaped curve = better characterization of variation

Figure 13: Bimodal histogram, not a bell shaped curve

Figure 14: Skewed histogram, not a bell shaped curve

Figure 15: Uniform histogram, not a bell shaped curve

Figure 16: Heavy-tailed histogram, not a bell shaped curve

Figure 17: Bell-shaped histogram, finally!

# Why concern yourself with the bell shaped curve?

- You can characterize individual observations
- You can characterize summary measures

Figure 18: Percentage within one standard deviation

Figure 19: Percentage within two standard deviations

Figure 20: Percentage within three standard deviations

#### Behavior of the mean versus an individual

- Central Limit Theorem
  - Sample mean is approximately normal
  - Even if individual observations are not
- Standard error:  $S/\sqrt{n}$

# Diagnosing distributional issues (1/2)

- For all data
  - $lacksquare ar{X}\gg X_{0.5}$
  - ullet  $ar{X}$  and/or  $X_{0.5}$  not midway between  $Q_1$  and  $Q_3$
  - ullet  $ar{X}$  and/or  $X_{0.5}$  not midway between min and max

# Diagnosing distributional issues (2/2)

- For non-negative data
  - $S>0.5 imes ar{X}$
- For data with an lower and/or upper bound
  - $Q_1$  = lower bound
  - $Q_3$  = upper bound
- Don't overdiagnose, especially with small sample sizes!

Figure 21: Lin et al 2022, PMID: 36126916

Figure 22: Excerpt from Table 1 of Lin et al 2022: ages

Figure 23: Excerpt from Table 1 of Lin et al 2022: CCI

Figure 24: Excerpt from Table 1 of Lin et al 2022: PHQ-2

Figure 25: Tosato et al 2021, PMID: 34352201

## Tosato 2021, PMID: 34352201 (continued)

Symptom persistence weeks after laboratory-confirmed severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) clearance is a relatively common long-term complication of Coronavirus disease 2019 (COVID-19). Little is known about this phenomenon in older adults. The present study aimed at determining the prevalence of persistent symptoms among older COVID-19 survivors and identifying symptom patterns.

# Tosato 2021, PMID: 34352201 (continued)

The mean age was  $73.1 \pm 6.2$  years (median 72, interquartile range 27), and 63 (38.4%) were women. The average time elapsed from hospital discharge was  $76.8 \pm 20.3$  days (range 25-109 days).

# Ielapi 2021, PMID: 34968328

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Figure 26: Ielapi et al 2021, PMID: 34968328

## Ielapi 2021, PMID: 34968328 (continued)

Background. Insomnia is one of the major health problems related with a decrease in quality of life (QOL) and also in poor functioning in night-shift nurses, that also may negatively affect patients' care. The aim of this study is to evaluate the prevalence of insomnia in night shift nurses.

## Ielapi 2021, PMID: 34968328 (continued)

Excerpt from Table 1. Data reported as mean ± standard deviation or median [Q1-Q3]

```
Overall (n = 2'355)
Age, years 40.4 ± 10.3
Months of work 168 [72-300]
Night shifts per month, number 6.3 ± 1.4
Time to reach workplace, minutes 45 [45-65]
Rest time, minutes 180 [4-240]
Rest in the afternoon, minutes 30 [0-120]
Number of coffees, mean 2.5 ± 1.5
Number of coffees during night shift, mean 1.4
± 1.1
```

## **Break**

- What have you just learned?
  - Computing the standard deviation
- What is coming next?
  - Visualization

Entries in the electronic health record by job title

Figure 27: Fastest computers by country

Figure 28: Demographic distribution of voters and non-voters in Texas

## Visualization

- Categorical data
  - Pie charts
  - Bar charts
- Continuous data
  - Bar charts
  - Error bars
  - Boxplot
  - Histogram
  - Plot all the data

Figure 29: Survivors among first class passengers

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Figure 30: Survivors among second class passengers

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Figure 31: Survivors among third class passengers

Figure 32: Bar chart showing proportion of passenger classes among deaths and survivors

Bar chart showing proportion of deaths and survivors among passenger classes

Bar chart showing only proportion of survivors among passenger classes

Bar chart showing proportion of survivors among passenger classes and sex

Bar chart showing proportion of survivors among sex and passenger classes

Figure 33: Bar chart showing average age among deaths and survivors

Bar chart with error bars showing proportion of survivors among sex and passenger classes

Figure 34: Boxplot showing ages of deaths and survivors

Figure 35: Annotated boxplot

Figure 36: Histogram of ages of passengers that died

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Figure 37: Histogram of ages of passengers that survived

Figure 38: SCatter plot of age vs survived

Figure 39: Scatter plot of age vs survived with jittering

Figure 40: Self reported versus observed adherence

Article feedback by type of article

Figure 41: Emergency care course test scores

Figure 42: Annual rainfall in selected cities

#### Which visualization to choose?

- Categorical data
  - Use a table
  - Minimize distance of key comparisons
- Continuous data
  - Small datasets: plot all the data
  - Large datasets: boxplots

Figure 43: Color combinations

## The RGB color system

- #rrggbb format
  - #000000 is pure black
  - #FFFFFF is pure white
  - #FF0000 is pure red
  - #00FF00 is pure green
  - #0000FF is pure blue
- You can mix and match to get 16,777,216 colors
  - #800080 is purple, #FF69B4 is pink, #40E0D0 is turquoise

Figure 44: Red plus green equals yellow

.

Figure 45: Red plus blue equals magenta

.

Figure 46: Green plus blue equals cyan

Figure 47: Yellow plus blue equals white

.

Figure 48: Magenta plus green equals white

.

Figure 49: Cyan plus red equals white

## The color cube

.

Figure 50: Illustration of the color cube

Figure 51: The red to green gradient on the color cube

# Rainbow

.

# The color cylinder

.

Figure 52: Color cylinder

Figure 53: Various foreground and background color combinations

Figure 54: A brighter version of the rainbow

## **Darker rainbow**

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Figure 55: A darker version of the rainbow

Figure 56: Color combinations using darker foregrounds and lighter backgrounds

Figure 57: DIffering luminance values of the rainbow

Figure 58: Rainbow colors on a white background

Figure 59: Rainbow colors on a black background

Figure 60: Rainbow colors showing a banding effect

Figure 61: One good set of color choices for nominal data

Examples of light to dark gradients

Figure 62: Examples of diverging gradients

#### **Color blindness**

- Up to 10% of your audience is color blind
  - Most common: red-green
- Suggestions
  - Use alternate cues (shape, shading)
  - Test your image
  - Find color blind friendly palettes.

Figure 63: Clothing mistake: using too many colors

Advertisement with a single red umbrella

Figure 64: Use of color to highlight a single individual

Figure 65: How many "5's" are in this figure?

Figure 66: Repeat question. How many "5's" are in this figure?

## Repeat quiz question 1

```
1 No Yes Total
2 Female 154 (33.3%) 308 (66.7%) 462 (100%
3 Male 709 (83.3%) 142 (16.7%) 851 (100%
4 Total 863 (65.7%) 450 (34.3%) 1313 (100%
```

This data table shows counts and ...

- 1. cell percents
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The problem with error bars is that they

- 1. fail to show if the data is skewed
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