# Not sure which Statistics course to take first? STAT 220, STAT 234, or the sequence STAT 244-245?

#### See a comparison of STAT 220 and 234:

statistics.uchicago.edu/~collins/introSTAT

#### Read a discussion of all of these courses:

collegecatalog.uchicago.edu/thecollege/statistics/#generalcourseinformation

# ...and feel free to drop by to talk with Dr. Linda Collins in Jones 205 (not available Tue/Thu afternoons)

#### Course prerequisites:

STAT 220: MATH 131 (or placement into MATH 151)

STAT 234: MATH 133, 153, or 162 (single-variable calculus) STAT 244: MATH 195, 200, or analysis (multi-variable calculus)

STAT 245: STAT 244 + linear algebra (STAT 243, MATH 20250 or higher)

#### Course website:

https://canvas.uchicago.edu/courses/32398

### Statistics Terminology

Like any field of inquiry, statistics assigns very specific meaning to some everyday words.

- ► sample (data), statistic
- ► population, parameter
- dataset: case, label, variable, value
- variable: quantitative, categorical
- distribution: variance, skew

#### Example: Wealth Distribution I

```
Rows: 19,674
Columns: 5
$ Y1984 <dbl> -169, 79470, 28660, NA, 5800, NA, NA, NA, ...
$ Y1989 <dbl> NA, 150500, 12300, NA, NA, NA, NA, 8685, NA, ...
$ Y1994 <dbl> NA, 145000, 3500, NA, 12000, NA, NA, NA, ...
$ Y2001 <dbl> NA, NA, 113050, 493000, NA, 41000, 148000...
$ Y2017 <dbl> NA, NA, NA, NA, 2000, 83000, NA, 0, NA, 1...
```

This data is from a study conducted by the Institute for Social Research at the University of Michigan, following around 15000 individuals over several decades in terms of a number of income and wealth variables. We are only showing total wealth for several years.

https://psidonline.isr.umich.edu/

Terms: popn vs. sample, cases vs. labels, variables vs. values

Variables: quantitative vs. categorical

#### Example: Wealth Distribution II

What is the distribution of Wealth in 2017?

A sample (or population) distribution of a variable has two parts:

- the set of values observed in the sample (or all values possible to observe in the population)
- 2. the relative frequency of occurrence for those values

#### Example: Wealth Distribution III

- ☐ A categorical variable places each case into one of several groups, or categories.
- ☐ A quantitative variable takes numerical values for which arithmetic operations such as adding and averaging make sense.
- ☐ The **distribution** of a variable tells us the values that a variable takes and how often it takes each value.

#### Example: Wealth Distribution IV

#### head(Wealth)

```
# A tibble: 5 x 5
  Y1984
         Y1989 Y1994
                        Y2001 Y2017
  <dbl>
         <dbl>
               <dbl>
                       <dbl> <dbl>
   -169
            NA
                    NA
                           NA
                                  NA
 79470 150500
               145000
                           NA
                                 NA
  28660
        12300
                  3500 113050
                                 NA
4
     NA
            NA
                    NA 493000
                                 NA
   5800
            NΑ
                 12000
                           NA
                                2000
```

#### tail(Wealth)

```
# A tibble: 5 \times 5
  Y1984 Y1989
                 Y1994 Y2001
                                Y2017
  <dbl> <dbl>
                 <dbl> <dbl>
                                <dbl>
     NA
          -500
                172000 95000
                               136001
1
2
3
4
5
     NA
             NA
                            NA
                                    NA
                      0
     NA
            NA
                     NA
                            NA
                                  1000
     NA
            NA
                    NA
                               149000
                            NA
     NA
             NA
                     NA
                            NA
                                81000
```

#### Example: Wealth Distribution V

Why all the NA's?

Do these data constitute a sample or a population?

### Example: Wealth Distribution VI

What is the distribution of family wealth in 2017?

Qualitatively describing the distribution of a quantitative variable: center, spread, and shape

summary(Wealth\$Y2017)

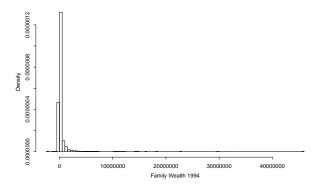
```
Min. 1st Qu. Median Mean 3rd Qu. Max. -2320000 285 28000 240328 162000 45745000 NA's 10067
```

Median= x: 50% of data is less than x. Q1= x: 25% of the data is less than x.

p-th percentile= x: p-proportion of the data less than x.

#### Example: Wealth Distribution VII

```
hist(Wealth$Y2017, freq=FALSE, xlab="Family Wealth 1994", breaks=100, main="")
```

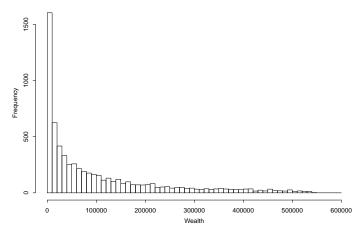


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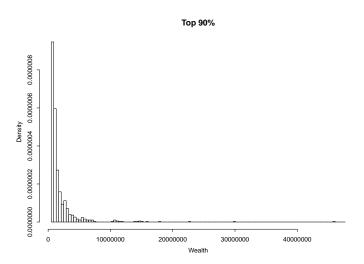
#### Example: Wealth Distribution VIII

Distribution of wealth for bottom 90% and top 10%

Bottom 90%

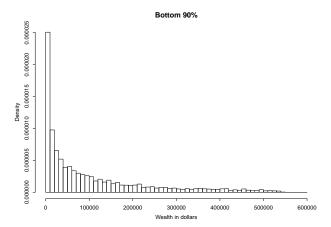


#### Example: Wealth Distribution 1>

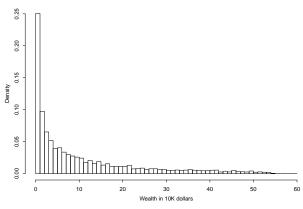


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#### Density histogram: Density of bar $\times$ width = percentage







# Software Installation: RStudio (and R) I

 $\begin{aligned} &\mathsf{RStudio} = \mathsf{the} \; \mathsf{work} \; \mathsf{environment} \\ &\mathsf{R} = \mathsf{the} \; \mathsf{engine} \; \mathsf{(a \; statistical \; programming \; language)} \end{aligned}$ 

To use the R code suggested for homework, you should install the moisaic package **once** at the start of the quarter.

```
install.packages("mosaic", ...and other packages)
```

Then, every time you start RStudio, type

require(mosaic)

### Example: Bicycle weight and commuting time I

#### glimpse(myBikeCommute)

Thanks to Dr. Jeremy Groves for providing his personal data.

http://www.bmj.com/content/341/bmj.c6801 Groves, J. Bicycle weight and commuting time: randomised trial, *British Medical Journal*, BMJ 2010;341:c6801.

# Example: Bicycle weight and commuting time II

#### head(myBikeCommute)

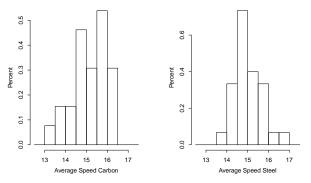
	Dino Davo	DIDUGITOO	HIHUCES	Avgspeed	TopSpeed	Month
	Steel 20/01/10	27.20	115.1	14.10	31.50	1Jan
2 Carbon	arbon 21/01/10	27.46	115.6	14.25	30.64	1Jan
3 Steel	Steel 25/01/10	27.20	115.8	14.10	30.92	1Jan
4 Carbon	arbon 26/01/10	27.52	113.9	14.49	33.02	1Jan
5 Carbon	arbon 27/01/10	27.51	119.2	13.84	30.92	2Feb
6 Steel	Steel 01/02/10	27.17	108.7	14.99	32.09	2Feb
7 Steel	Steel 03/02/10	27.16	117.7	13.84	32.09	2Feb
8 Carbon	arbon 03/02/10	27.49	123.3	13.37	29.58	2Feb
9 Carbon	arbon 08/02/10	27.48	112.5	14.65	34.02	2Feb
10 Steel	Steel 09/02/10	27.09	112.6	14.43	32.71	2Feb
11 Carbon	arbon 11/02/10	27.44	117.7	13.99	32.00	3Mar
12 Carbon	arbon 01/03/10	27.49	108.6	15.18	32.71	3Mar
13 Carbon	arbon 03/03/10	27.49	110.9	14.82	34.71	3Mar
4 Carbon 5 Carbon 6 Steel 7 Steel 8 Carbon 9 Carbon 10 Steel 11 Carbon 12 Carbon	arbon 26/01/10 arbon 27/01/10 Steel 01/02/10 Steel 03/02/10 arbon 08/02/10 arbon 08/02/10 Steel 09/02/10 arbon 11/02/10 arbon 01/03/10	27.52 27.51 27.17 27.16 27.49 27.48 27.09 27.44 27.49	113.9 119.2 108.7 117.7 123.3 112.5 112.6 117.7 108.6	14.49 13.84 14.99 13.84 13.37 14.65 14.43 13.99 15.18	33.02 30.92 32.09 32.09 29.58 34.02 32.71 32.00 32.71	2F 2F 2F 2F 2F 2F 3M 3M

Why not alternating Steel, Carbon, Steel, Carbon, Steel, etc.?

Terms: popn vs. sample, cases vs. labels, variables vs. values

Variables: quantitative vs. categorical

### Example: Bicycle weight and commuting time III



Compare speed distributions: center, spread, shape Steel: same average?, less spread, right skewed Carbon: same average?, more spread, left skewed

# Summarizing a distribution with a center: Average I

by(myBikeCommute\$AvgSpeed,myBikeCommute\$Bike,mean)

myBikeCommute\$Bike: Carbon

[1] 15.19

\_\_\_\_\_

myBikeCommute\$Bike: Steel

[1] 15.04

Average of average speed is about the same for both frame types.

mean(myBikeCommute\$Distance)

[1] 27.16

The average distance is close to claimed distance: 27 miles

#### **Definition:**

sample average 
$$= \overline{x} =$$
 "x-bar"  $= \frac{1}{n} \sum_{i=1}^{n} x_i$ 

n = sample size

Reminder: The average is the balancing point of the data

For **any** sample of size 
$$n$$
,  $\sum_{i=1}^{n} (x_i - \overline{x}) = 0$ 

Start on the left side: 
$$\sum_{i=1}^{n} (x_i - \overline{x})$$

# Summarizing a distribution with a center: Average III

$$\sum_{i=1}^{n} (x_i - \overline{x}) = \text{rewritten expression}$$

$$= \text{rewritten again}$$

$$= \text{and rewritten again}$$

$$= \text{until arriving at the right side} = 0$$

$$\sum_{i=1}^{n} x_i - n\overline{x} = \dots$$

#### Summarizing a distribution with a center: Median

Median: Splits the data in half, half above half below.

by(myBikeCommute\$Distance,myBikeCommute\$Bike,median)

```
myBikeCommute$Bike: Carbon
```

[1] 27.38

-----

myBikeCommute\$Bike: Steel

[1] 27.01

median(myBikeCommute\$Distance)

[1] 27.19

The median distance is close to claimed distance: 27 miles

sort(mvBikeCommute\$Distance)

```
 \begin{array}{c} [1] & 25.86 & 26.60 & 26.74 & 26.88 & 26.90 & 26.91 & 26.91 & 26.91 & 26.92 \\ [10] & 26.94 & 26.94 & 26.94 & 26.95 & 26.99 & 27.00 & 27.00 & 27.01 & 27.02 \\ [19] & 27.02 & 27.03 & 27.05 & 27.06 & 27.09 & 27.10 & 27.16 & 27.16 \\ [28] & 27.17 & 27.20 & 27.20 & 27.27 & 27.29 & 27.31 & 27.31 & 27.32 & 27.32 \\ [37] & 27.33 & 27.34 & 27.34 & 27.36 & 27.38 & 27.39 & 27.40 & 27.40 \\ [46] & 27.43 & 27.44 & 27.45 & 27.46 & 27.48 & 27.49 & 27.49 & 27.49 & 27.51 \\ [55] & 27.52 & 27.52 \end{array}
```

### Measuring Spread of Data Distribution

The average devation  $\frac{1}{n} \sum_{i=1}^{n} (x_i - \overline{x})$  always = 0! So, we need a different measure for distance (spread)

There are many measures of spread:

- mean squared deviation (MSD or "variance"),
- ▶ mean absolute deviation (MAD),
- ▶ standard deviation (SD) = root  $MSD = RMSD = \sqrt{MSD}$ ,
- ▶ interquartile range (IQR= range of middle 50% of data)
- range,
- ...and more (not covered in this course).

#### Loss Functions I

No matter which one number we might choose to measure center, we are summarizing an entire distribution with one number.

- There is a cost.
- We lose information.
- ▶ We should measure that loss and be aware of its magnitude.
- Statisticians measure loss numerically with a loss function
- A loss function measures the distance of the data from the one-number summary (the "center").

A loss function is a measure of distance (spread).

#### Loss Functions II

#### Let's consider two common loss functions

► The sum (or mean) of absolute deviations:

$$SAD(w) = \sum_{i=1}^{n} |x_i - w|$$
  $MAD(w) = \frac{1}{n} \sum_{i=1}^{n} |x_i - w|$ 

► The sum (or mean) of squared deviations:

$$SSD(w) = \sum_{i=1}^{n} (x_i - w)^2$$
  $MSD(w) = \frac{1}{n} \sum_{i=1}^{n} (x_i - w)^2$ 

#### Loss Functions III

What value of w should we choose if loss is SAD? If SSD?

It seems reasonable that w should be in the "center" of the data for each measure. But which value in the middle would be best?

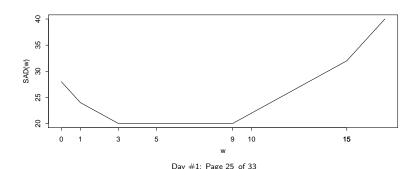
One optimality criteria: Choose w that minimizes SAD or SSD.

### What is so special about the median?

Consider the data  $x_1 = 9, x_2 = 3, x_3 = 15, x_4 = 1$ 

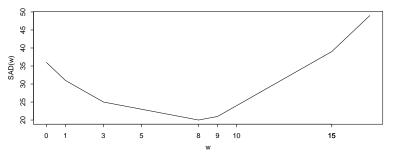
What does the SAD(w) function look like for these data?

```
ww=seq(0,17,.1)
SAD <- function(w) { sum( abs(x-w) ) }
plot(ww,sapply(ww, SAD), type=c('l'), xlab="w", ylab="SAD(w)")
axis(1,at=x)</pre>
```



# What is so special about the median? II

```
 y <- c(9,3,15,8,1) \\ SAD1=function(w)\{sum(abs(y-w))\} \\ plot(ww,sapply(ww, SAD1), type=c('l'), xlab="w", ylab="SAD(w)") \\ axis(1,at=y)
```

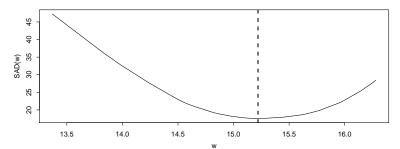


Where is the function SAD(w) smallest (minimized)?

### What is so special about the median? II

Looking at the data: Carbon frame AvgSpeed

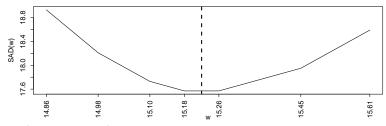
What does the SAD(w) function look like for these data?



Where is the function SAD(w) smallest (minimized)?

# What is so special about the median? IV

#### Zooming in:



sort(carbonSpeed)

```
[1] 13.37 13.84 13.99 14.25 14.49 14.54 14.58 14.65 14.82 [10] 14.86 14.98 15.10 15.18 15.26 15.45 15.61 15.64 15.75 [19] 15.78 15.95 15.96 15.99 16.15 16.15 16.25 16.28
```

median(carbonSpeed)

[1] 15.22

# What is so special about the average? I

Consider again the data  $x_1 = 9, x_2 = 3, x_3 = 15, x_4 = 1$ 

What is the SSD(w) function for these data?

$$\sum_{i=1}^{4} (x_i - w)^2 = (9 - w)^2 + (3 - w)^2 + (15 - w)^2 + (1 - w)^2$$
$$= 4w^2 - 56w + 316$$

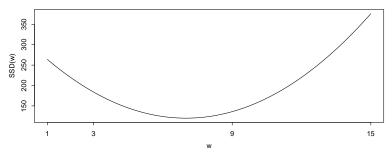
So, as ugly as 
$$\sum_{i=1}^{n} (x_i - w)^2$$
 originally looks

it's just a smooth quadratic function (convex, opening up).

# What is so special about the average? II

What does the SSD(w) function look like for these data?

SSD <- function(w) 
$$\{ sum((x-w)^2) \}$$



What value w minimizes  $SSD(w) = 4w^2 - 56w + 316$ ?  $\frac{d}{dw}SSD(w) = 8w - 56$ .

Set the derivative = 0 and solve for w: w = 7.

# What is so special about the average?

What value w minimizes SSD(w) for **any** sample:  $x_1, x_2, ..., x_n$ ?

We want to minimize the following function with respect to w:

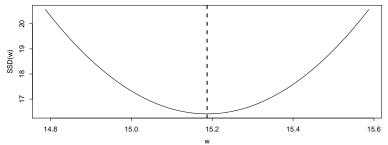
$$f(w) = SSD(w) = \sum_{i=1}^{n} (x_i - w)^2$$

**On your own:** Show that minimizer is  $w = \overline{x}$  (average). Then, check that the average is the *unique* minimum (not just one of several values that attain the minimum, as for the median).

We say that  $\overline{x}$  is a "least squares" statistic since it minimizes the sum of squared deviations.

# What is so special about the average? IV

Consider again the bike commute data: Carbon frame AvgSpeed What does the SSD(w) function look like for these data?



Where is the function SSD(w) smallest (minimized)?

mean(carbonSpeed)

[1] 15.19

### Formulas for Sample Average, Variance, SD

sample average 
$$= \overline{x} =$$
 "x-bar"  $= \frac{1}{n} \sum_{i=1}^{n} x_i$ 

sample variance 
$$= s^2 =$$
 "s-squared"  $= \frac{1}{n-1} \sum_{i=1}^{n} (x_i - \overline{x})^2$ 

sample standard deviation 
$$= s = \sqrt{s^2} = \sqrt{\frac{1}{n-1} \sum_{i=1}^{n} (x_i - \overline{x})^2}$$
  
= "typical" distance from the average

Why divide by (n-1) instead of n for sample variance and SD?