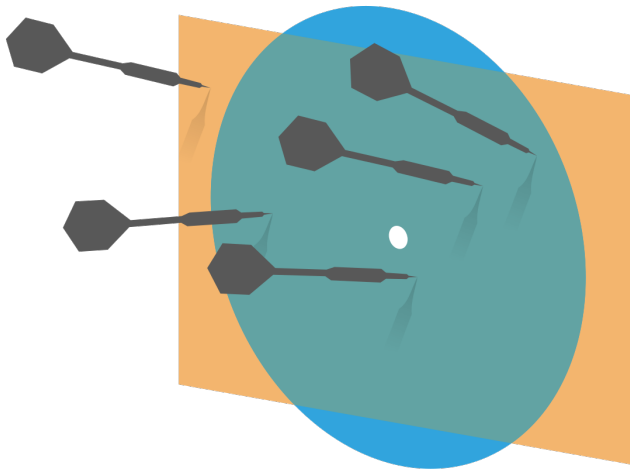


# Probability and Statistics for Computer Science



“The statement that “The average US family has 2.6 children” invites mockery” – Prof. Forsyth reminds us about critical thinking

Credit: wikipedia

# Last lecture

- ✱ Welcome/Orientation
- ✱ Big picture of the contents
- ✱ Lecture 1 - Data Visualization & Summary (I)
- ✱ **Orientation quiz due today**

# Warm up question:

- ✱ What kind of data is a letter grade?
- ✱ What do you ask for usually about the stats of an exam with numerical scores?

(1) A: Categorical      (2) write  
B: ordinal              in chat  
C: Continuous

# Objectives

✱ **Grasp Summary Statistics**

*3Ms*

*Mean Median Mode*

✱ **Learn more Data Visualization for Relationships**

*Std. Var. IQR*



# Summarizing 1D continuous data

For a data set  $\{x\}$  or annotated as  $\{x_i\}$ , we summarize with:

$N$  items

## \* Location Parameters

Mean ( $\mu$ ), Median, Mode

## \* Scale parameters

Standard deviation ( $\sigma$ ),  
Variance ( $\sigma^2$ )

Interquartile range

# Summarizing 1D continuous data

## ✱ Mean

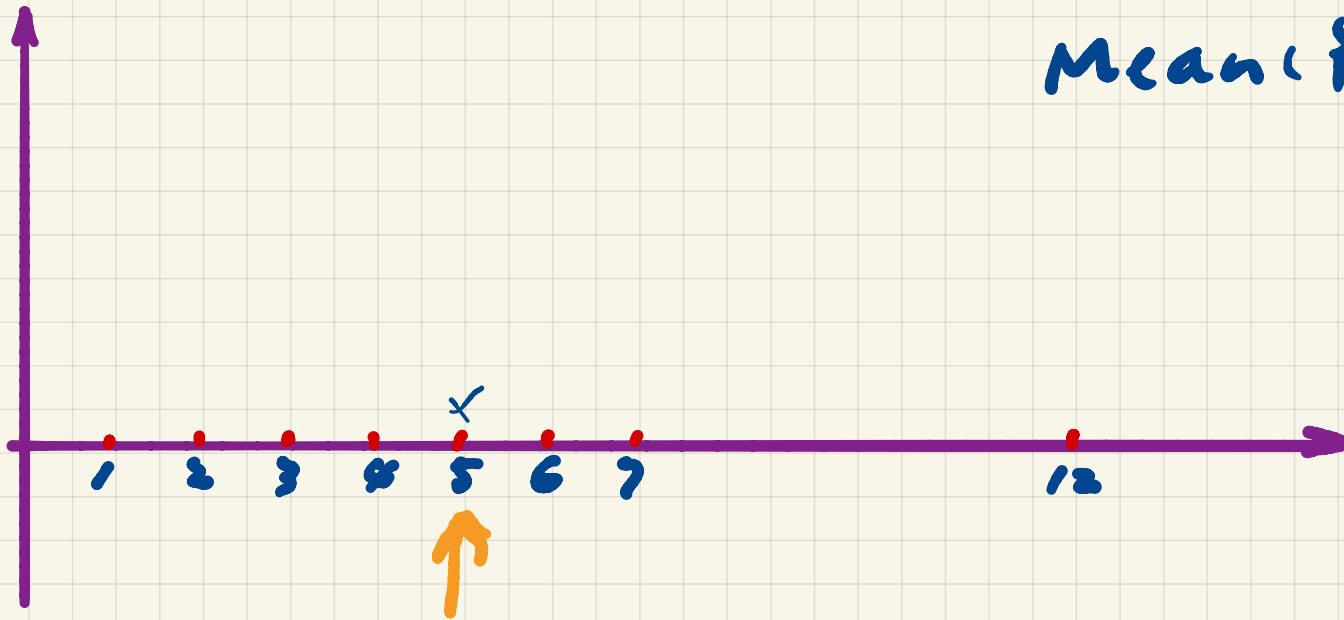
$$\text{mean}(\{x_i\}) = \frac{1}{N} \sum_{i=1}^N x_i$$

It's the centroid of the data geometrically, by identifying the data set at that point, you find the center of balance.

$\{x_i\} \quad i \in [1, 8]$

$\{x_i\} = 1, 2, 3, 4, 5, 6, 7, 12$

$\text{Mean}(\{x_i\}) = 5$



# Properties of the mean

- ✱ Scaling data scales the mean

$$\begin{aligned} & \text{mean}(\{kx_i + c\}) \\ &= k \cdot \text{mean}(\{x_i\}) + c \\ \text{mean}(\{k \cdot x_i\}) &= k \cdot \text{mean}(\{x_i\}) + c \end{aligned}$$

- ✱ Translating the data translates the mean

$$\text{mean}(\{x_i + c\}) = \text{mean}(\{x_i\}) + c$$

# Less obvious properties of the mean

- ✱ The signed distances from the mean

sum to 0

$$\sum_{i=1}^N (x_i - \text{mean}(\{x_i\})) = 0$$

- ✱ The mean minimizes the sum of the squared distance from any real value

$$\underset{\mu}{\text{argmin}} \sum_{i=1}^N (x_i - \mu)^2 = \text{mean}(\{x_i\})$$

$$\text{Proof: } \sum_{i=1}^N (x_i - \text{mean}\{x_i\}) = 0$$

$$\begin{aligned} \text{LHS} &= \sum_{i=1}^N x_i - \sum_{i=1}^N \text{mean}\{x_i\} \quad \rightarrow \text{Const} \\ &= \sum_{i=1}^N (x_i - N \cdot \text{mean}\{x_i\}) \\ &= \sum_{i=1}^N (x_i - N \cdot \frac{\sum_{i=1}^N x_i}{N}) \\ &= \sum_{i=1}^N (x_i - \sum_{i=1}^N x_i) = 0 \end{aligned}$$

$$\text{Proof: } \underset{\mu}{\text{Argmin}} \left( \sum_{i=1}^N (x_i - \mu)^2 \right) = \text{mean}(\{x_i\})$$

$\underset{\mu}{\text{Argmin}}$ : Argument  $\mu$  that minimizes the function that follows

LHS =  $\hat{\mu}$   $\rightarrow$  the special  $\mu$  that minimizes  $f(\mu) = \sum_{i=1}^N (x_i - \mu)^2$

To find  $\hat{\mu}$ , set  $\frac{df(\mu)}{d\mu} = 0$  & solve it

One way is to use the Chain rule

$$f(\mu) = \sum_{i=1}^N h(\mu) = \sum_{i=1}^N g^2(\mu) \quad g = x_i - \mu$$

$$\frac{df}{d\mu} = \frac{d \sum h}{d\mu} = \sum \frac{dh}{d\mu} = \sum_{i=1}^N \frac{dh}{dg} \cdot \frac{dg}{d\mu}$$

Proof:  $\underset{\mu}{\operatorname{Argmin}} \left( \sum_{i=1}^N (x_i - \mu)^2 \right) = \operatorname{mean}(\{x_i\})$

$$\frac{df(\mu)}{d\mu} = \sum \frac{dh}{dg} \frac{dg}{d\mu} = \sum 2g \cdot (-1) = 0$$

$$h = g^2$$

$$g = x_i - \mu$$

$$\Rightarrow \sum_{i=1}^N g = 0$$

$$\Rightarrow \sum_{i=1}^N (x_i - \mu) = 0$$

$$\sum_{i=1}^N x_i - N \cdot \mu = 0$$

$$\therefore \hat{\mu} = \frac{\sum_{i=1}^N x_i}{N} = \operatorname{mean}(\{x_i\})$$

$$\frac{d^2 f(\mu)}{d\mu^2} ?$$



Q1:

✪ What is the answer for

$mean(\{mean(\{x_i\})\})$  ?

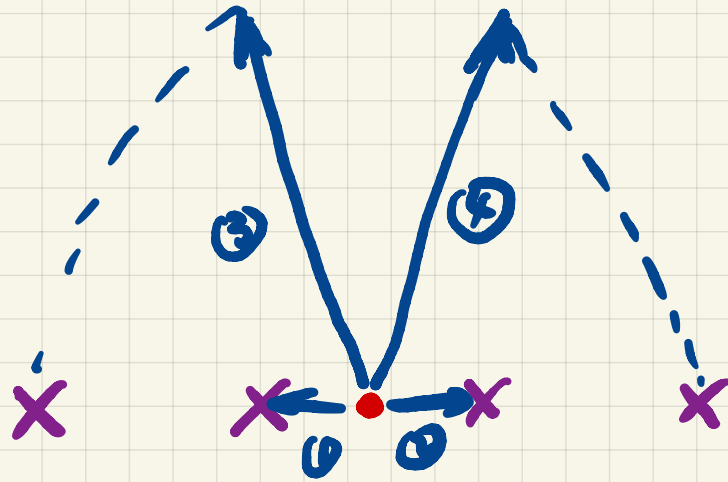
A.  $mean(\{x_i\})$     B. unsure    C. 0

# Standard Deviation ( $\sigma$ )

✱ The standard deviation

$$\begin{aligned} \text{std}(\{x_i\}) &= \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \text{mean}(\{x_i\}))^2} \\ &= \sqrt{\text{mean}(\{(x_i - \text{mean}(\{x_i\}))^2\})} \end{aligned}$$

How much the  
data spreads  
out wrt mean



$$\text{std} = \sqrt{\frac{1}{4} \sum_{i=1}^4 d_i^2}$$

Q2. Can a standard deviation of a dataset be -1?

A. YES

B. NO



# Properties of the standard deviation

- ✱ Scaling data scales the standard deviation

$$\text{std}(\{k \cdot x_i\}) = |k| \cdot \text{std}(\{x_i\})$$

↑ ↑

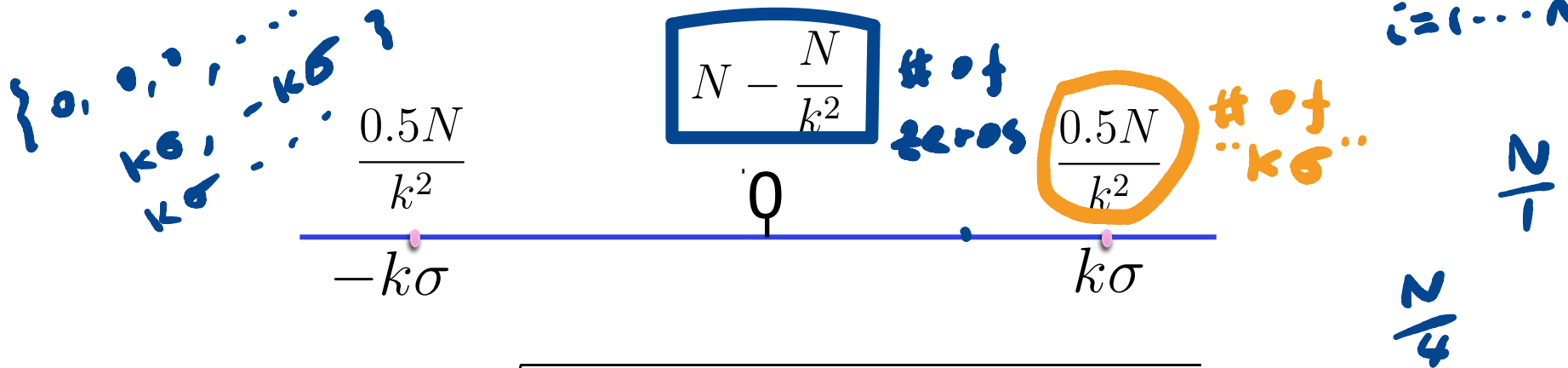
- ✱ Translating the data does **NOT** change the standard deviation

$$\text{std}(\{x_i + c\}) = \text{std}(\{x_i\})$$

# Standard deviation: Chebyshev's inequality (1<sup>st</sup> look)

✱ At most  $\frac{N}{k^2}$  items are  $k$  standard deviations ( $\sigma$ ) away from the mean
 
 $N$  is # of items in  $\{x_i\}_{i=1 \dots N}$

✱ Rough justification: Assume mean = 0



$$std = \sqrt{\frac{1}{N} \left[ \left( N - \frac{N}{k^2} \right) 0^2 + \frac{N}{k^2} (k\sigma)^2 \right]} = \sigma$$

# Variance ( $\sigma^2$ )

✱ Variance = (standard deviation)<sup>2</sup>

$$\text{var}(\{x_i\}) = \frac{1}{N} \sum_{i=1}^N (x_i - \text{mean}(\{x_i\}))^2$$

✱ Scaling and translating similar to standard

deviation  $\text{var}(\{k \cdot x_i\}) = k^2 \cdot \text{var}(\{x_i\})$

$$\text{var}(\{x_i + c\}) = \text{var}(\{x_i\})$$

## Q3: Standard deviation

✱ What is the value of  
 $std(\{mean(\{x_i\})\})$  ?

A. 0    B. 1    C. unsure



# Standard Coordinates/normalized data

- ✱ The *mean* tells where the data set is and the *standard deviation* tells how spread out it is. If we are interested only in comparing the shape, we could

define:

$$\hat{x}_i = \frac{x_i - \text{mean}(\{x_i\})}{\text{std}(\{x_i\})}$$

for every  $i$

- ✱ We say  $\{\hat{x}_i\}$  is in standard coordinates

# Q4: Mean of standard coordinates

✱  $\text{mean}(\{\hat{x}_i\})$  is:

↓ A. 1  B. 0 C. unsure

$\mu$

$$\hat{x}_i = \frac{x_i - \text{mean}(\{x_i\})}{\text{std}(\{x_i\})}$$

# Q5: Standard deviation ( $\sigma$ ) of standard coordinates

✱ Std( $\{\hat{x}_i\}$ ) is:

A. 1 B. 0 C. unsure

$$\text{std}(\{k x_i\}) = |k| \text{std}(\{x_i\})$$

↓  
6

$$\hat{x}_i = \frac{x_i - \text{mean}(\{x_i\})}{\text{std}(\{x_i\})}$$

# Q6: Variance of standard coordinates

✱ Variance of  $\{\hat{x}_i\}$  is:

A. 1   B. 0   C. unsure

$\sigma^2$

$$\hat{x}_i = \frac{x_i - \text{mean}(\{x_i\})}{\text{std}(\{x_i\})}$$

# Q7: Estimate the range of data in standard coordinates

✱ Estimate as close as possible, 90% data is within:

A. [-10, 10]

B. [-100, 100]

C. [-1, 1]

D. [-4, 4]

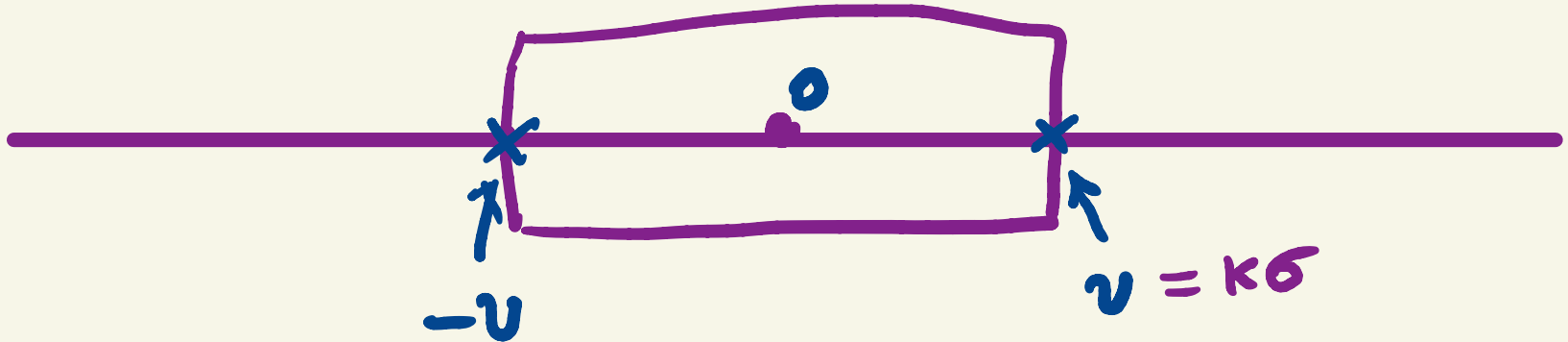
E. others

$$\hat{x}_i = \frac{x_i - \text{mean}(\{x_i\})}{\text{std}(\{x_i\})}$$

$$\frac{\frac{N}{k^2}}{N} = \frac{1}{k^2} \leq 10\%$$

$\{\hat{x}_i\}$

$\geq 90\%$



$$k\sigma = k$$

$$\therefore \sigma(\{\hat{x}_i\}) = 1$$

# Standard Coordinates/normalized data to $\mu=0, \sigma=1, \sigma^2=1$

- ✱ Data in standard coordinates always has  
mean = 0; standard deviation = 1;  
variance = 1.
- ✱ Such data is unit-less, plots based on this  
sometimes are more comparable
- ✱ We see such normalization very often in  
statistics

# Median

- ✱ We first sort the data set  $\{x_i\}$
- ✱ Then *if* the number of items  $N$  is **odd**  
median = middle item's value  
*if* the number of items  $N$  is **even**  
median = mean of middle 2 items' values



# Properties of Median

- ✱ Scaling data scales the median

$$\mathit{median}(\{k \cdot x_i\}) = k \cdot \mathit{median}(\{x_i\})$$

- ✱ Translating data translates the median

$$\mathit{median}(\{x_i + c\}) = \mathit{median}(\{x_i\}) + c$$

# Percentile

- ✱  $k^{\text{th}}$  percentile is the value relative to which  $k\%$  of the data items have smaller or equal numbers
- ✱ Median is roughly the  $50^{\text{th}}$  percentile

$\{1, 2, 3, 4, 5, 6, 7, 12\}$   
75<sup>th</sup> percentile = ? 6  $\neq 0.75$   
75%

# Interquartile range

✱  $iqr = (75\text{th percentile}) - (25\text{th percentile}) > 0$

✱ Scaling data scales the interquartile range

$$iqr(\{k \cdot x_i\}) = |k| \cdot iqr(\{x_i\})$$

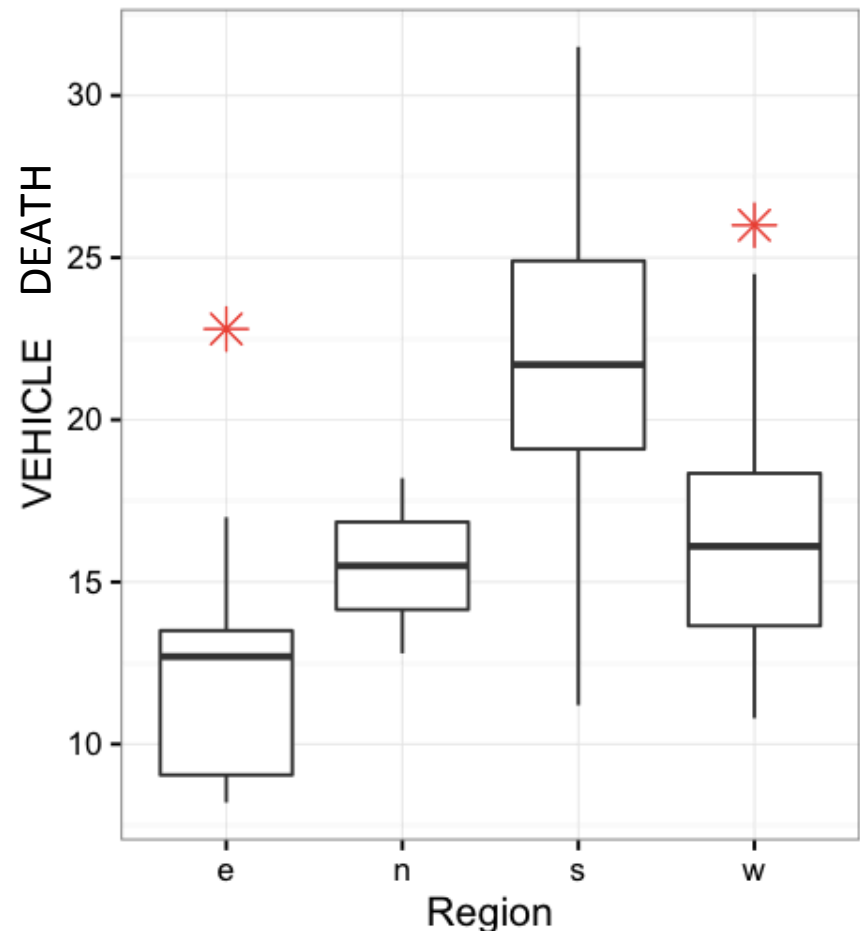
✱ Translating data does **NOT** change the interquartile range

$$iqr(\{x_i + c\}) = iqr(\{x_i\})$$

# Box plots

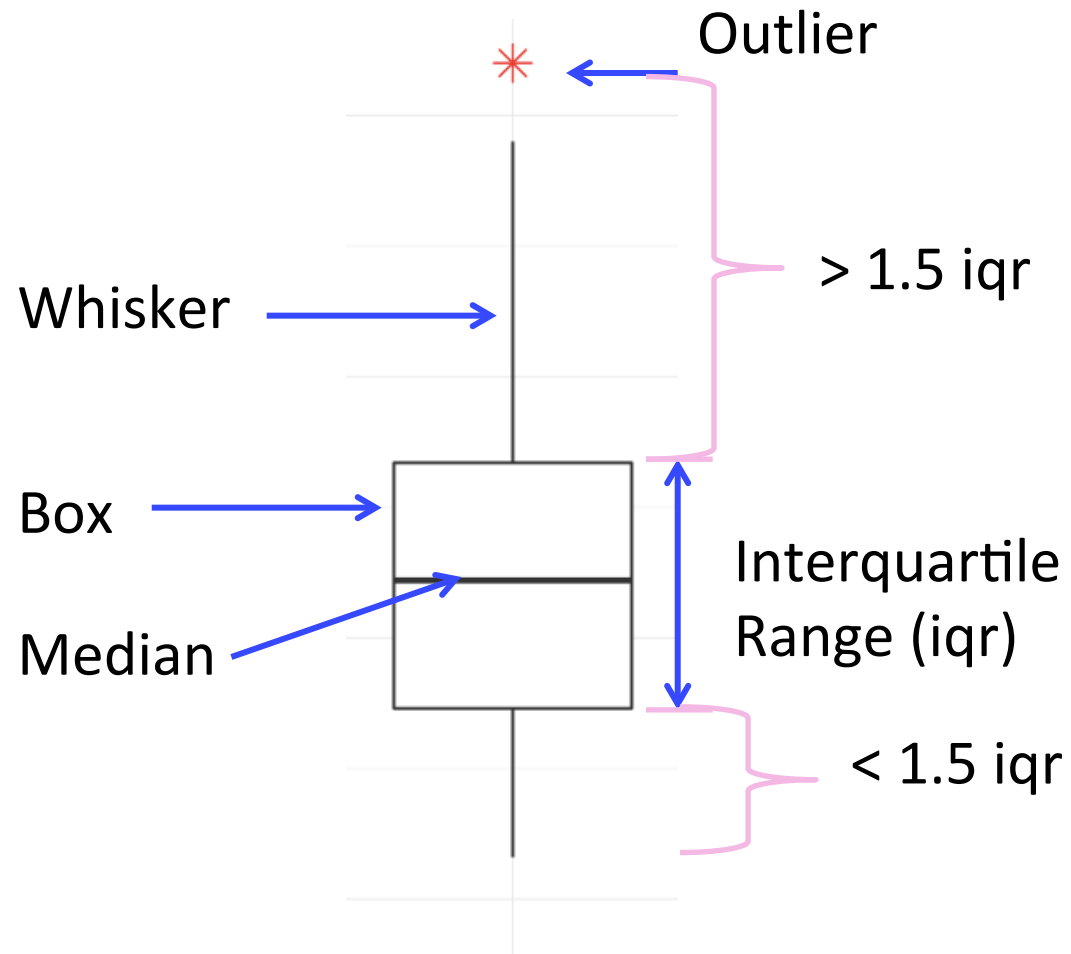
- ✱ Boxplots
- ✱ Simpler than histogram
- ✱ Good for outliers
- ✱ Easier to use for comparison

## Vehicle death by region



# Boxplots details, outliers

✱ How to  
define  
outliers?  
(the default)



## Q. TRUE or FALSE

mean is more sensitive to outliers than median

A. True

B. False

## Q. TRUE or FALSE

interquartile range is more sensitive to outliers than std.

A. True

B. False

# Sensitivity of summary statistics to outliers

- ✱ mean and standard deviation are very sensitive to outliers
- ✱ median and interquartile range are not sensitive to outliers



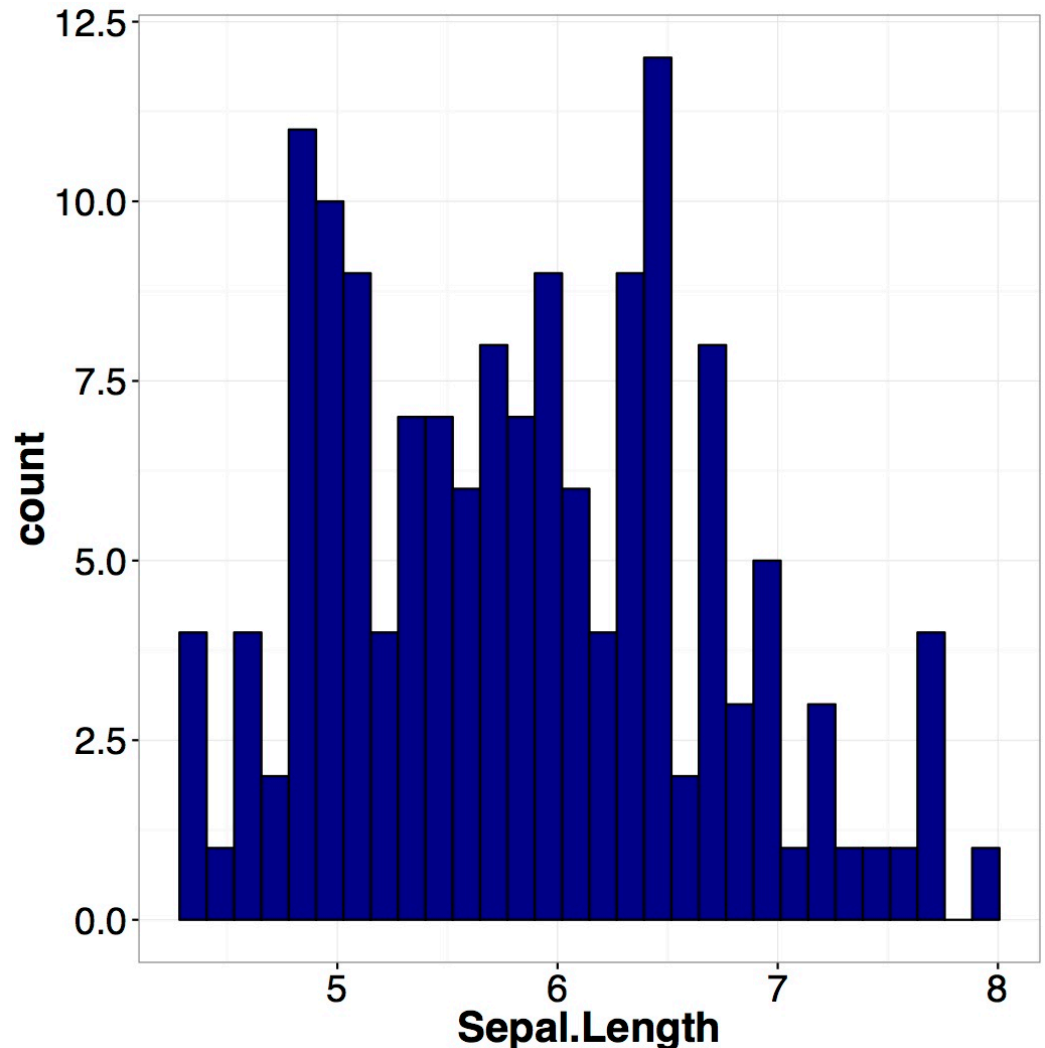
# Modes

- ✱ Modes are peaks in a histogram
- ✱ If there are more than 1 mode, we should be curious as to why

# Multiple modes

✱ We have seen the “iris” data which looks to have several peaks

Data: “iris” in R

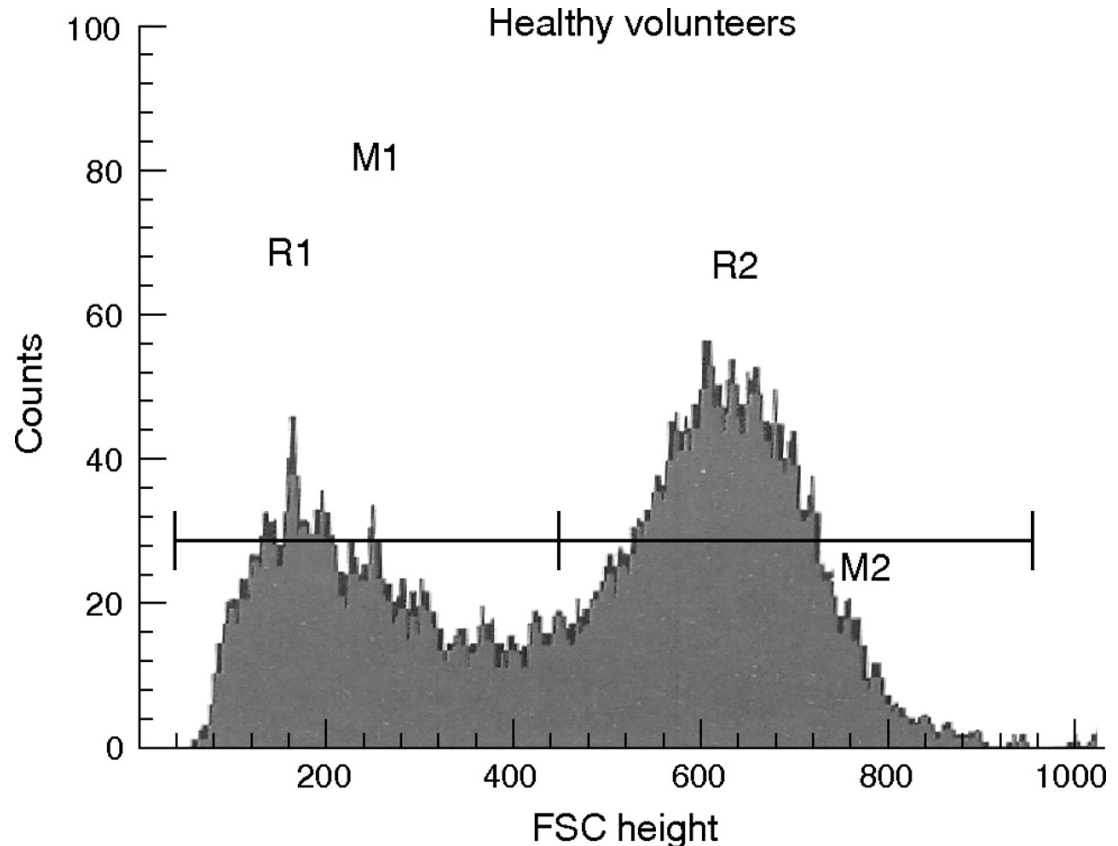


# Example Bi-modes distribution

- ✱ Modes may indicate multiple populations

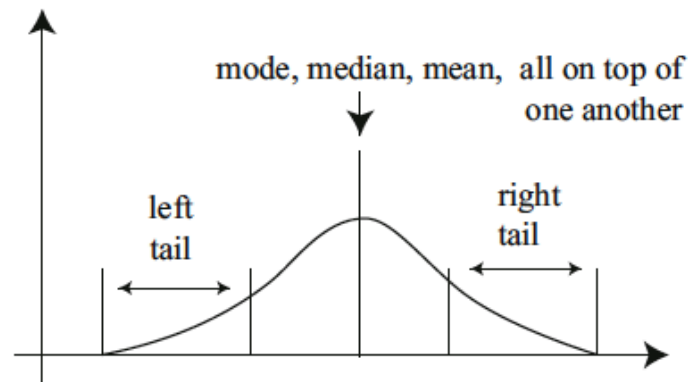
Data: Erythrocyte cells in healthy humans

Piagnerelli, JCP 2007

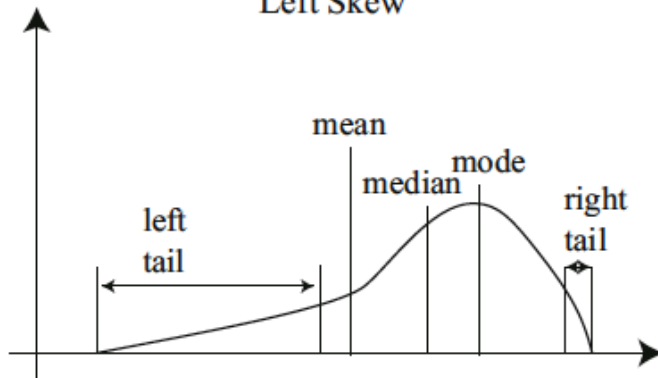


# Tails and Skews

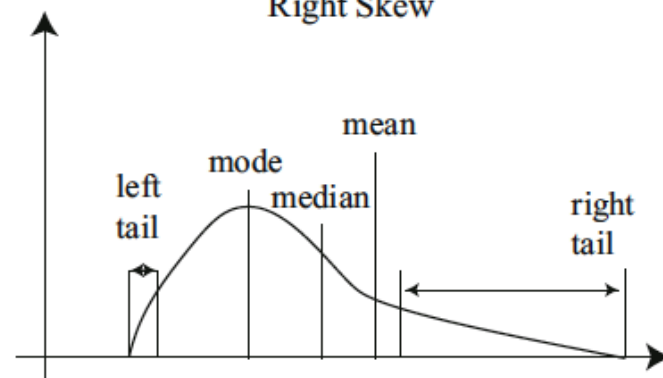
Symmetric Histogram



Left Skew

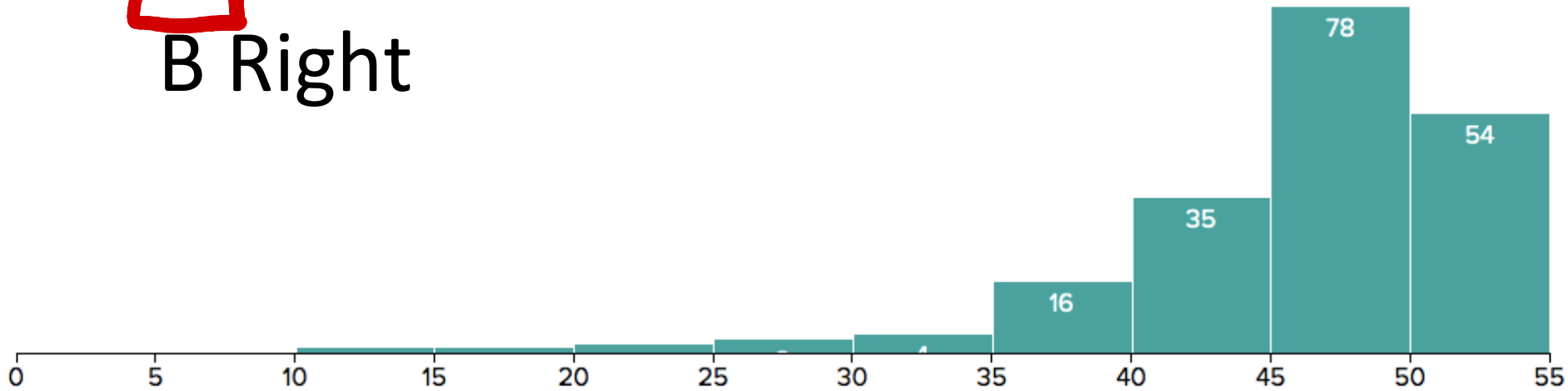


Right Skew



# Q. How is this skewed?

**A** Left  
B Right



Median = 47

*mean = 46*  
*mean < median  $\Rightarrow$  left*

# Assignments

- ✱ **HW1** due Thurs. Feb. 4.
- ✱ **Quiz 1 (open 4:30pm today until Mon. next week)**
- ✱ Reading upto Chapter 2.1
- ✱ Next time: the quantitative part of correlation coefficient

# Additional References

- ✱ Charles M. Grinstead and J. Laurie Snell  
"Introduction to Probability"
- ✱ Morris H. Degroot and Mark J. Schervish  
"Probability and Statistics"

See you next time

*See  
You!*

