

# Week 2: Examining Numerical Data

Professor Kathryn Jacobs  
Today's music theme: More 80's hits!



# Quantitative Variables: General

“Numerical variables”

Mathy math

Measured

Not all numbers will  
count

Name	Company	Company _Number	Serving	Calories	Fat	Sodium	Carbs	Fiber	Sugars	Protein
AppleJacks	K	2	1.00	117	.6	143	27	.5	15.0	1.0
Boo Berry	G	1	1.00	118	.8	211	27	.1	14.0	1.0
Cap'n Crunch	Q	3	.75	144	2.1	269	31	1.1	16.0	1.3
Cinnamon Toast Crunch	G	1	.75	169	4.4	408	32	1.7	13.3	2.7
Cocoa Blasts	Q	3	1.00	130	1.2	135	29	.8	16.0	1.0
Cocoa Puffs	G	1	1.00	117	1.0	171	26	.8	14.0	1.0
Cookie Crisp	G	1	1.00	117	.9	178	26	.5	13.0	1.0
Corn Flakes	K	2	1.00	101	.1	202	24	.8	3.0	2.0
Corn Pops	K	2	1.00	117	.2	120	28	.3	15.0	1.0
Crispix	K	2	1.00	113	.3	229	26	.1	3.0	2.0
Crunchy Bran	Q	3	.75	120	1.3	309	31	6.4	8.0	1.3
Froot Loops	K	2	1.00	118	.9	150	26	.8	12.0	2.0
Frosted Mini-Wheats	K	2	1.00	175	.8	5	41	5.0	10.0	5.0
Golden Grahams	G	1	.75	149	1.3	359	33	1.3	14.7	2.7
Honey Nut Clusters	G	1	1.00	214	2.7	249	46	2.8	17.0	4.0
Honey Nut Heaven	Q	3	1.00	192	3.7	216	38	3.5	13.0	4.0
King Vitaman	Q	3	1.50	80	.7	173	17	.9	4.0	1.3
Kix	G	1	1.30	87	.5	205	20	.8	2.3	1.5
Life	Q	3	.75	160	1.9	219	33	2.7	8.0	4.0

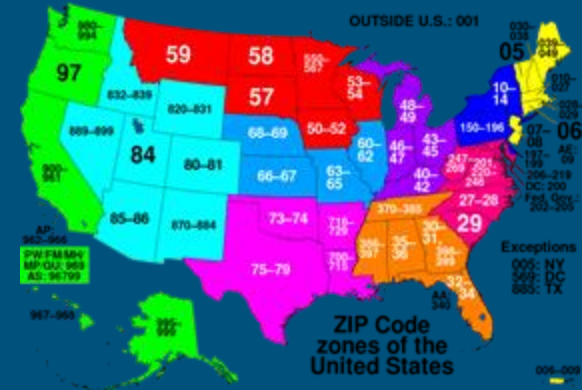
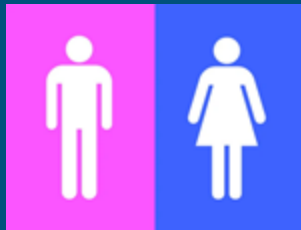
# Quantitative Variables: General

Numerical variables that are NOT quantitative:

Zip codes

Football jersey numbers

Male = 1, Female = 0



# Quantitative Variables: THINK

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3 examples of numerical variables that ARE quantitative

3 examples of numerical variables that ARE NOT quantitative



# Quantitative Variables: PAIR

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3 examples of numerical variables that ARE quantitative

3 examples of numerical variables that ARE NOT quantitative



# Quantitative Variables: SHARE

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3 examples of numerical variables that ARE quantitative

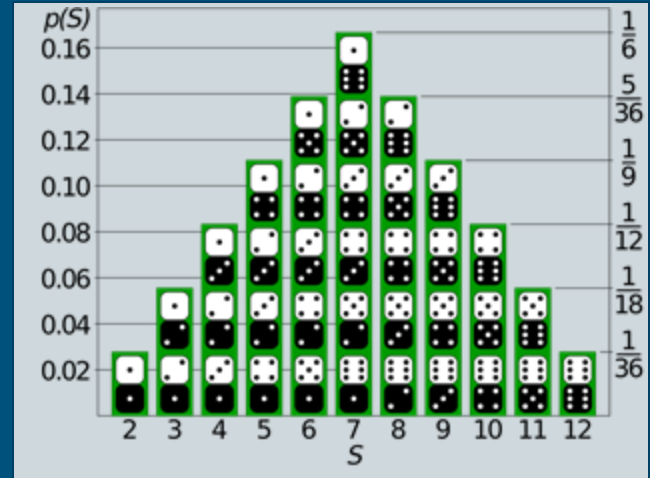
3 examples of numerical variables that ARE NOT quantitative



# Distributions

What is a distribution?

“How likely different values are for a given variable”



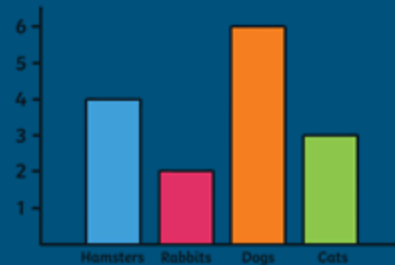
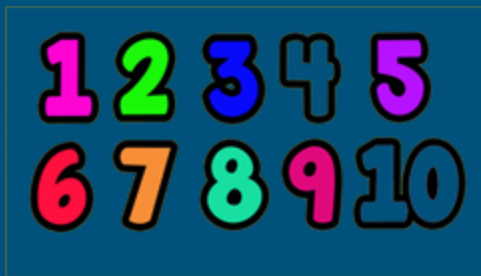
How are different values *distributed* in the population

We can find this information using graphs, plots, and number summaries

# To Do

## Summary Statistics

- Center: mean, median, mode
- Spread: standard deviation, range, IQR
- Percentiles
- 5 number summary



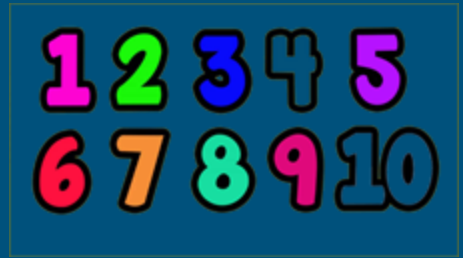
## Visualization

- Dot plots
- Frequency tables
- Stem-and-leaf plots
- Histograms
- Box plots



# Summary Statistics

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Center

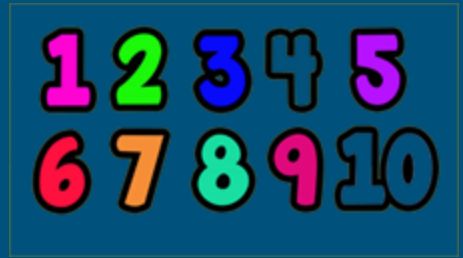
Spread

Percentiles

5- number summary

# Summary Statistics

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Center

Spread

Percentiles

5- number summary

# Measures of center: Mean, Median, Mode

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Mean: mathematical average

Median: center number if numbers are ordered smallest-largest

Mode: most common number



# Measures of center: Mean

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Parameter:  $\mu$  (mew)

Statistic:  $\bar{x}$

Add all numbers together, divide by  
total sample size/population

Population

$$\mu = \frac{\sum x}{N}$$

Sample

$$\bar{x} = \frac{\sum x}{n}$$

# Measures of center: Median

Parameter:  $\eta$  (eta)

Statistic:

The image shows the symbol for the sample mean, which is a capital letter X with a tilde (~) over it, enclosed in a white square box.

1, 3, 3, **6**, 7, 8, 9

Median = **6**

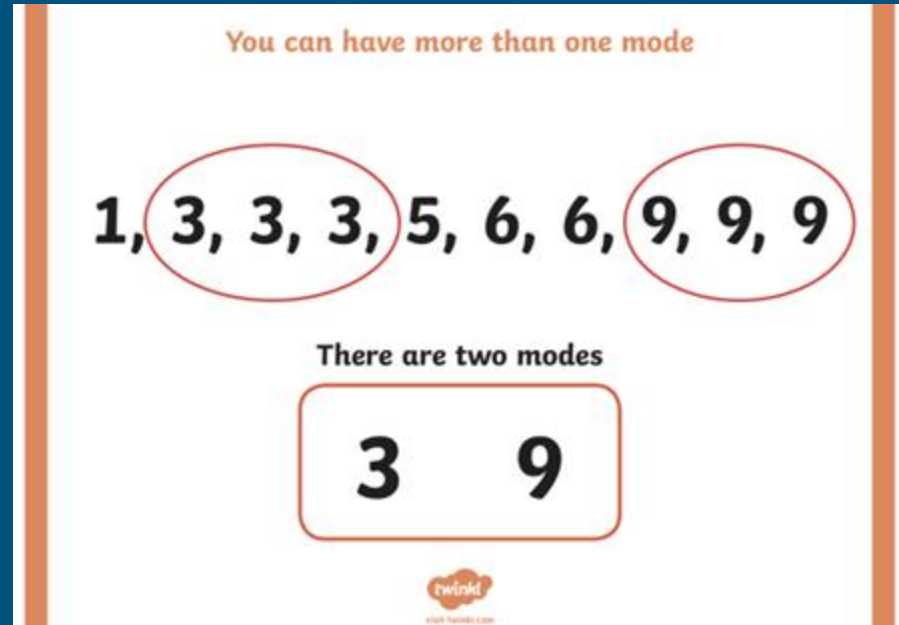
1, 2, 3, **4**, **5**, 6, 8, 9

Median =  $(4 + 5) \div 2$   
= **4.5**

# Measures of center: Mode

No symbol, we just say mode

Uniform, unimodal, bimodal,  
multimodal



# Measures of center: Resistance

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Is mean, median, or mode best?

Depends on our data!

Some things, like extreme values, will affect some measures more than others



# Summary Statistics

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Center

Spread

Percentiles

5- number summary



# Spread: General

Spread describes how *varied* our distribution is

Are most of our values close together, or spread out?

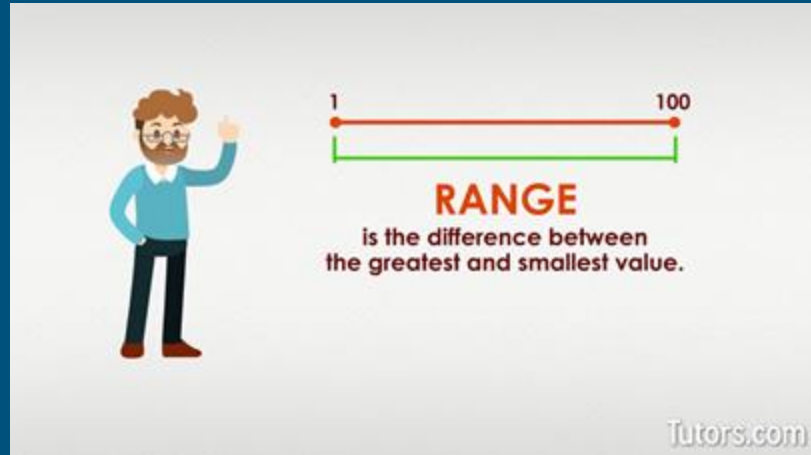


Name	Height	Weight
Timothy Leary	5' 10" (180 cm)	175 lbs (79 kg)
Jeffrey Pardo	5' 10" (180 cm)	175 lbs (79 kg)
Jeffrey Pardo	5' 10" (180 cm)	175 lbs (79 kg)
Charles Austin	5' 10" (180 cm)	175 lbs (79 kg)
James Brown	5' 10" (180 cm)	175 lbs (79 kg)
Greg Neri	5' 10" (180 cm)	175 lbs (79 kg)

# Spread: Range

Very simple!

Largest value - smallest value



# Spread: Standard deviation

Representation of just how varied a variable is: “average distance of a data point from the mean”

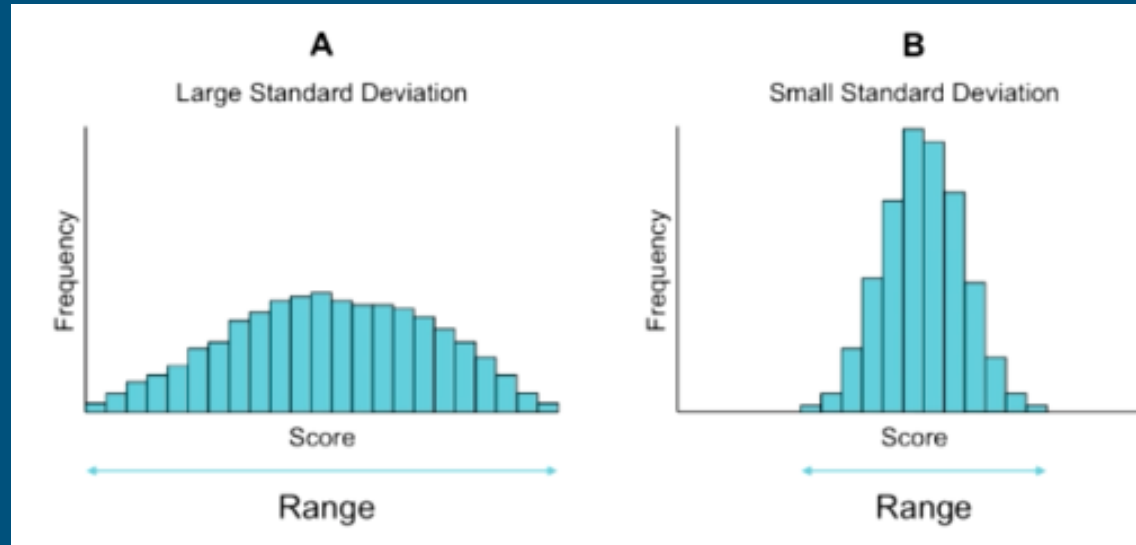
Large SD:

extremely varied data

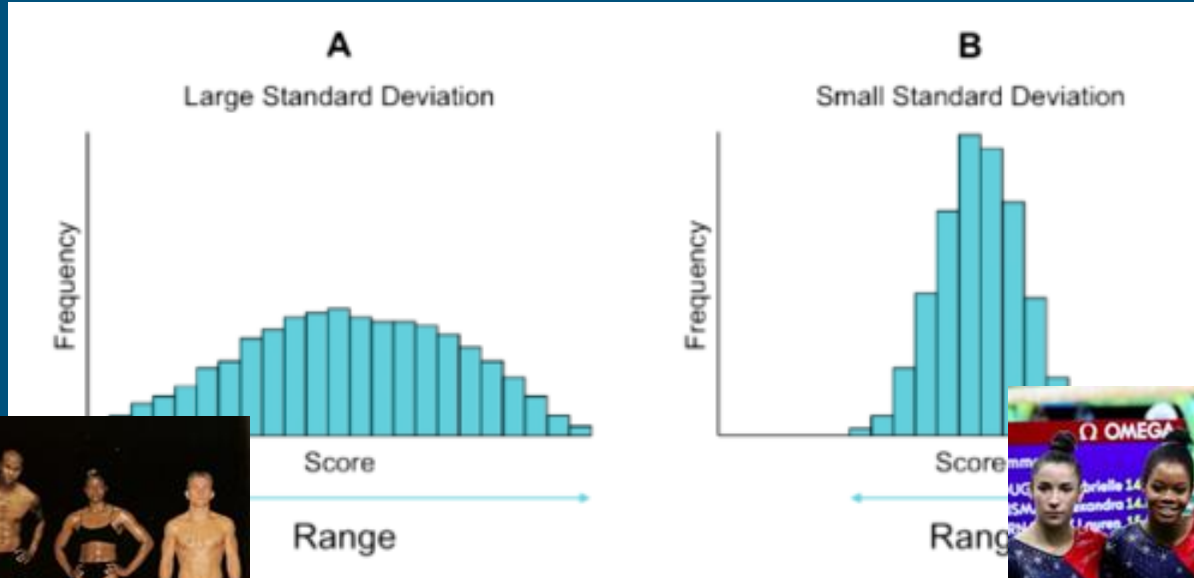
Small SD:

Data all clusters

close to mean



# Spread: Standard deviation



# Spread: Standard deviation

Population	Sample
$\sigma = \sqrt{\frac{\sum(x_i - \mu)^2}{n}}$ <p><math>\mu</math> - Population Average <math>x_i</math> - Individual Population Value <math>n</math> - Total Number of Population</p>	$S = \sqrt{\frac{\sum(x_i - \bar{x})^2}{n-1}}$ <p><math>\bar{x}</math> - Sample Average <math>x_i</math> - Individual Population Value <math>n</math> - Total Number of Sample</p>

# Standard deviation: Practice

Let's calculate the standard deviation of this data set:

[ 1, 1, 3, 5, 5 ]

Mean = 3

How extreme is the value of 5?

Sample
$S = \sqrt{\frac{\sum (x_i - \bar{x})^2}{n-1}}$
<p><b>X</b> - Sample Average <b>x<sub>i</sub></b> - Individual Population Value <b>n</b> - Total Number of Sample</p>

# Spread: Variance

Variance = standard deviation squared

Represents TOTAL amount of variation within a data set

Doesn't mean much by itself- used in formulas for other things

Population	Sample
$\sigma^2 = \frac{\sum(x_i - \mu)^2}{n}$ <p><math>\mu</math> - Population Average <math>x_i</math> - Individual Population Value <math>n</math> - Total Number of Population <math>\sigma^2</math> - Variance of Population</p>	$S^2 = \frac{\sum(x_i - \bar{x})^2}{n-1}$ <p><math>\bar{x}</math> - Sample Average <math>x_i</math> - Individual Population Value <math>n</math> - Total Number of Sample <math>S^2</math> - Variance of Sample</p>

# Notation: for reference

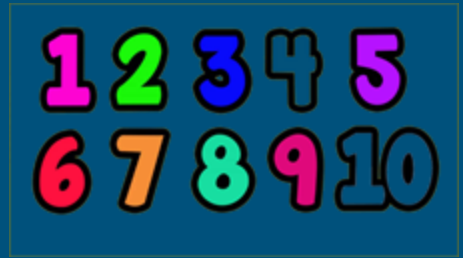
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Name	Population Parameters	Sample Statistics
Mean	$\mu$	$\overline{X}$
Median	$\eta$	$\tilde{X}$
Mode	No symbol	No symbol
Range	R	R
Variance	$\sigma^2$	$s^2$
Standard Deviation	$\sigma$	s
Sample Size	N	n
Estimates	$\hat{\sigma}$	n/a



# Summary Statistics

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Center

Spread

Percentiles

5- number summary

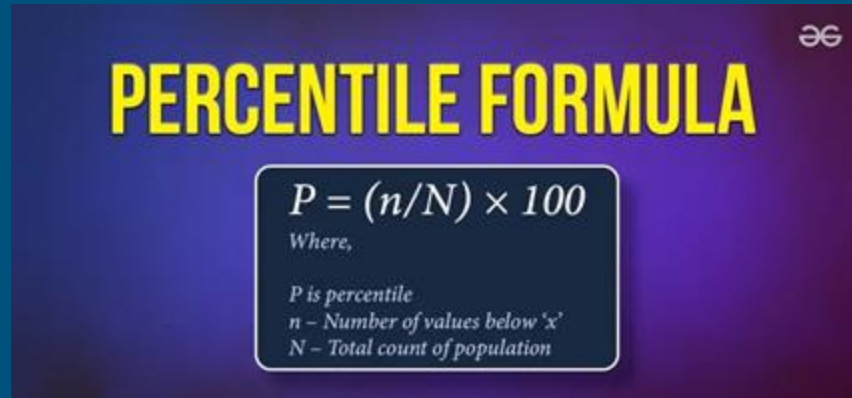
# Percentiles

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Compare a single value to the entire data set

“This number is bigger than 80% of the rest of the data”

Median = 50th percentile



**PERCENTILE FORMULA**

$$P = (n/N) \times 100$$

Where,

- P* is percentile
- n* – Number of values below 'x'
- N* – Total count of population

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# Percentiles: Example

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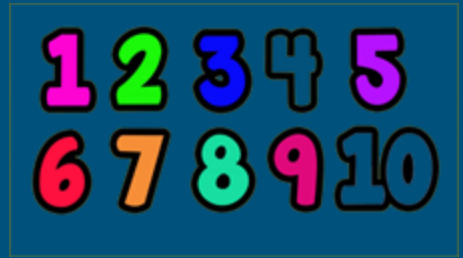
For the following sample, what percentile is a value of 4?

[3, 9, 4, 5, 5, 8, 2]

1. Order smallest-largest
2. Count values smaller than target number
3. Divide by total sample size

# Summary Statistics

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Center

Spread

Percentiles

5- number summary

# 5-number summary

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## 5 Number Summary

Min = Smallest number

Q1 = Median of the first half of the data

Q2 = Median

Q3 = Median of the second half of the data

Max = Largest number

# 5-number summary

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1. Order numbers smallest - largest
2. Find min and max
3. Find median
4. (numbers below median) median (numbers above median)
5. Find Q1 and Q3 using numbers in (parentheses)
6. [min, Q1, median, Q3, max]

IQR = Interquartile range =  $Q3 - Q1$

# 5-number summary: IQR

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Another way to measure spread of distribution

$Q3 - Q1$

# Practice

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Find the mean, median, mode, 5-number summary, and IQR for the following data set:

[ 3, 5, 7, 7, 2, 4, 2, 2, 8, 6, 5, 7, 7, 7, 4]

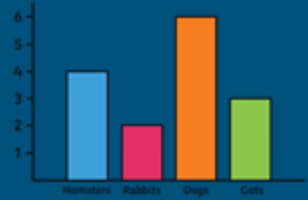
What number is at the 80th percentile?

[ 2, 2, 2, 3, 4, 4, 5, 5, 6, 7, 7, 7, 7, 7, 8]



# Visualization

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Dot plots

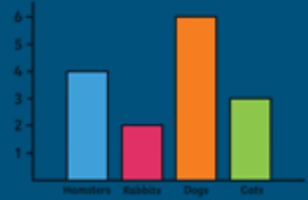
Frequency tables

Stem-and-leaf plots

Histograms

Box plots

# Visualization



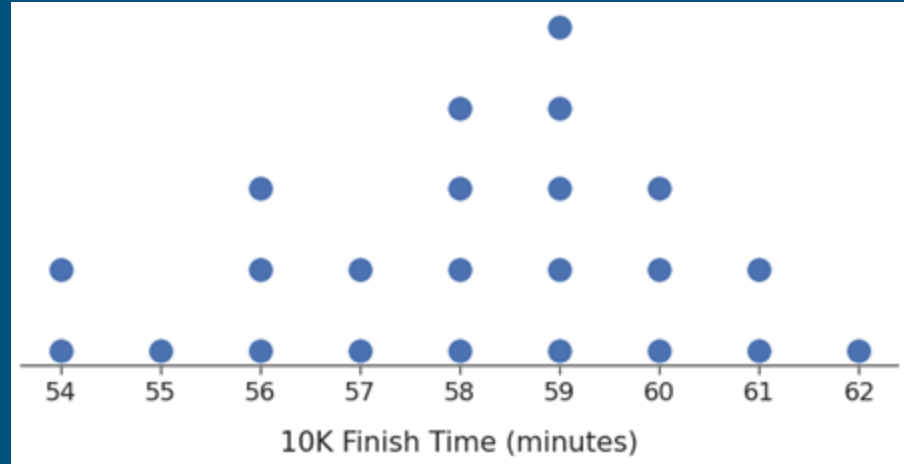
Dot plots

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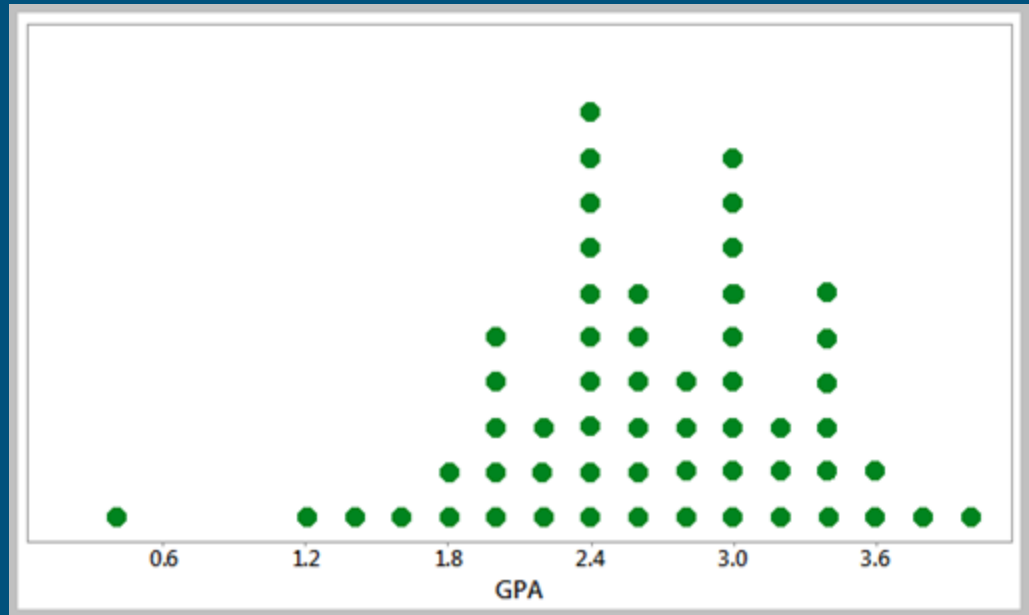
# Dot Plots

Each dot represents 1 case

Only discrete data

Can be used to approximate mode(s)

Does this distribution have a mode?



# Visualization

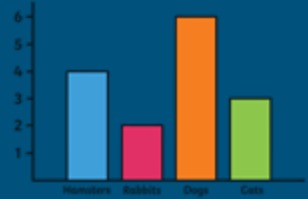
Dot plots

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Score	Frequency
6	2
7	3
8	7
9	7
10	1

# Frequency tables

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Let's construct a frequency table

5, 7, 3, 10, 18, 10, 10, 5, 13, 13, 18

# Frequency tables: Grouped data

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What if we don't have any repeat values, or have lots and lots of values?

<u>Age Group</u>	<u>Frequency</u>	<u>Percent</u>
21-25	87	43.1
26-30	43	21.3
31-35	25	12.3
16-20	17	8.4
36-40	15	7.4
46-50	4	2.0
51-55	4	2.0
41-45	3	1.5
56-60	3	1.5
<u>61+</u>	<u>1</u>	<u>0.5</u>

# Visualization



Dot plots

Frequency tables

Stem-and-leaf plots

Histograms

Box plots

1, 5, 12, 7, 22, 4, 5, 27, 3, 13, 19

Stem	Leaf
0	1 3 4 5 5 7
1	2 3 9
2	2 7

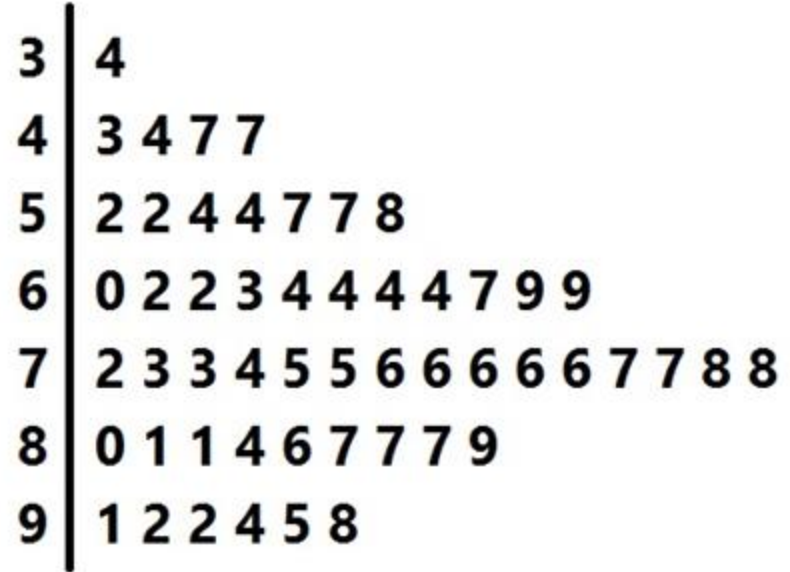
# Stem and Leaf plots

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Good for small data sets

Data ordered - usually smallest to largest

What is our mode here?





# Visualization

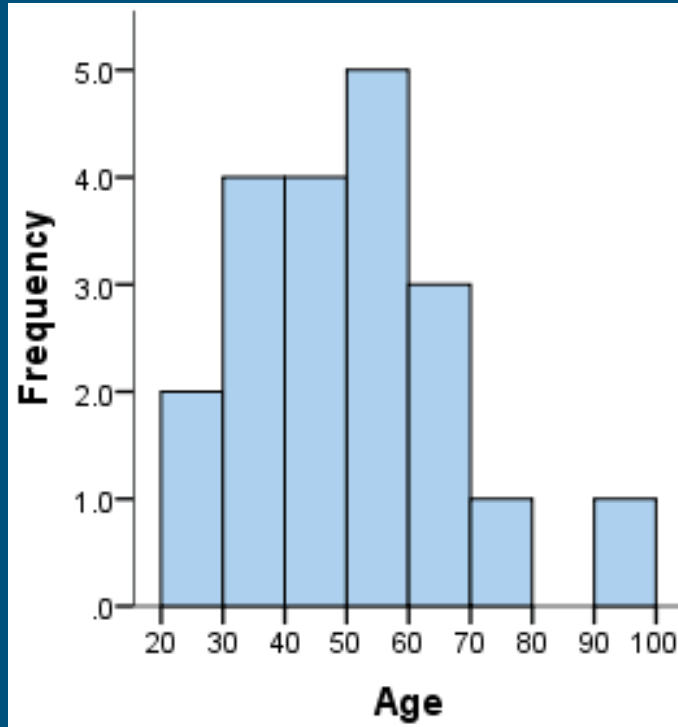
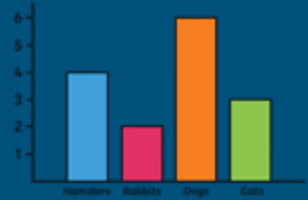
Dot plots

Frequency tables

Stem-and-leaf plots

Histograms

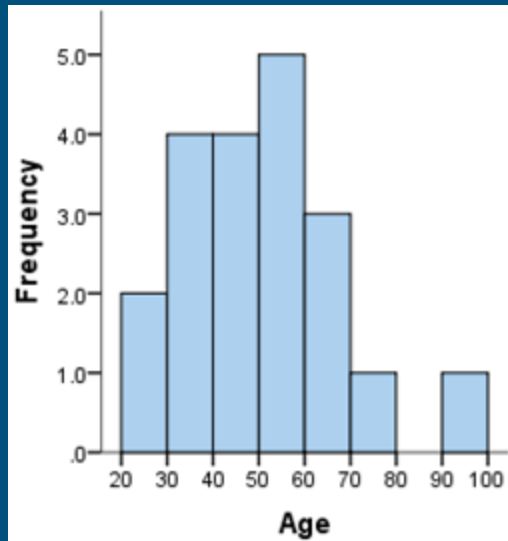
Box plots



# Histograms: General

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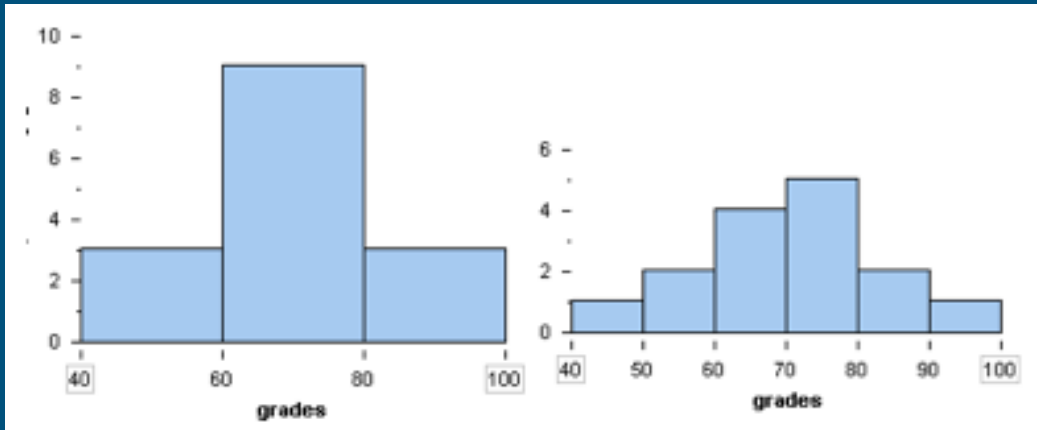
- Group continuous variables into ranges, or *bins*
- Frequency of cases in each bin is added up, and graphed along y axis



# Histograms: Bin Width

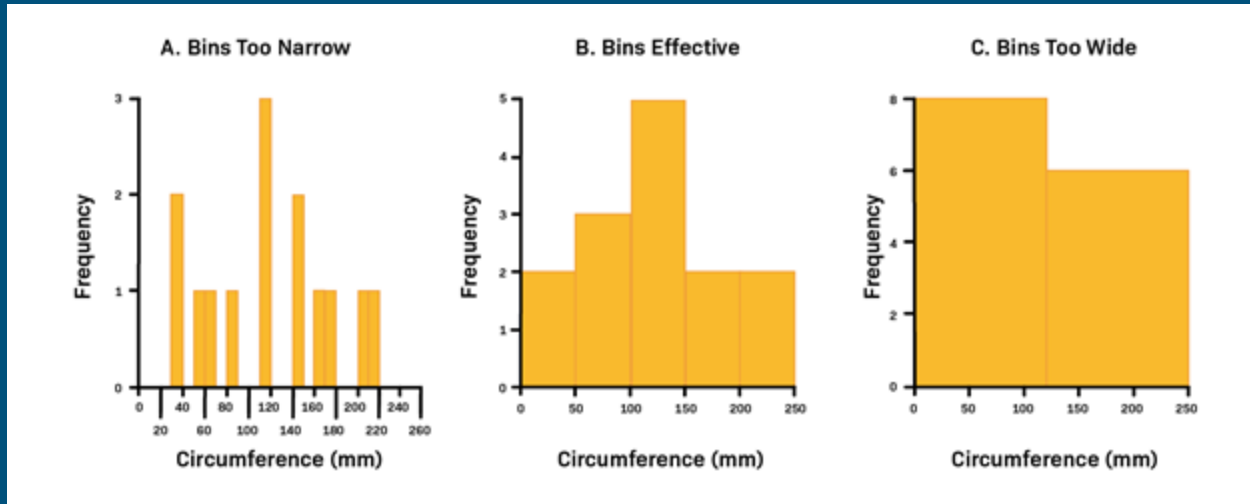
In general, more bins = better

Narrower bins give us more information



# Histograms: Bin Width

Sometimes if bins are too narrow, we lose information



# Histograms from: Frequency tables

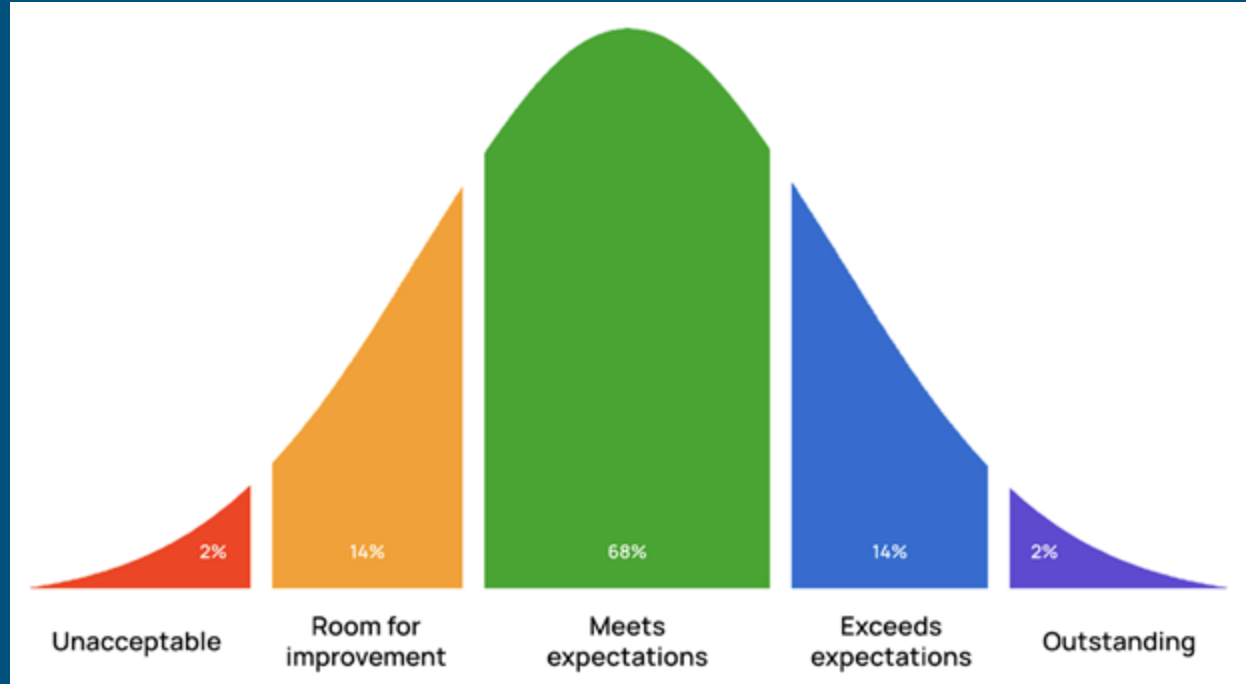
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Let's make a histogram!

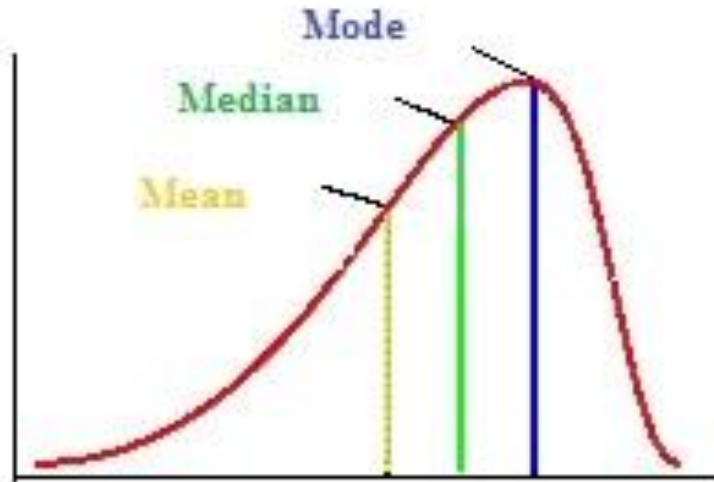
<u>Age Group</u>	<u>Frequency</u>	<u>Percent</u>
21-25	87	43.1
26-30	43	21.3
31-35	25	12.3
16-20	17	8.4
36-40	15	7.4
46-50	4	2.0
51-55	4	2.0
41-45	3	1.5
56-60	3	1.5
<u>61+</u>	<u>1</u>	<u>0.5</u>

# Histograms: Normal curve

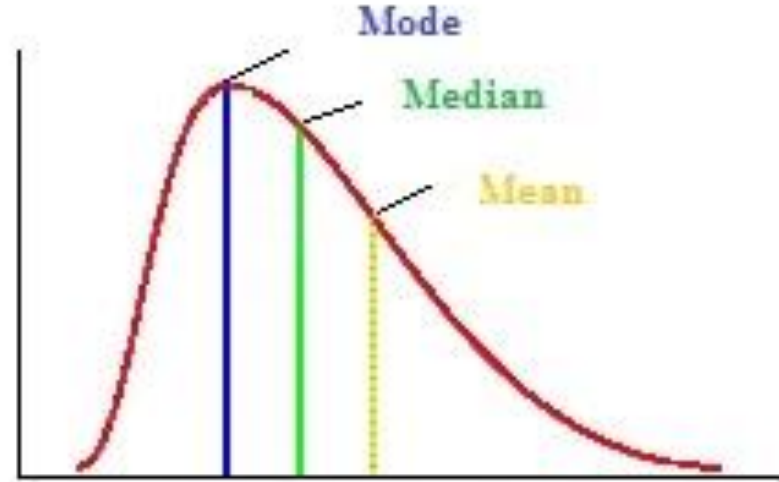
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# Histograms: Skew

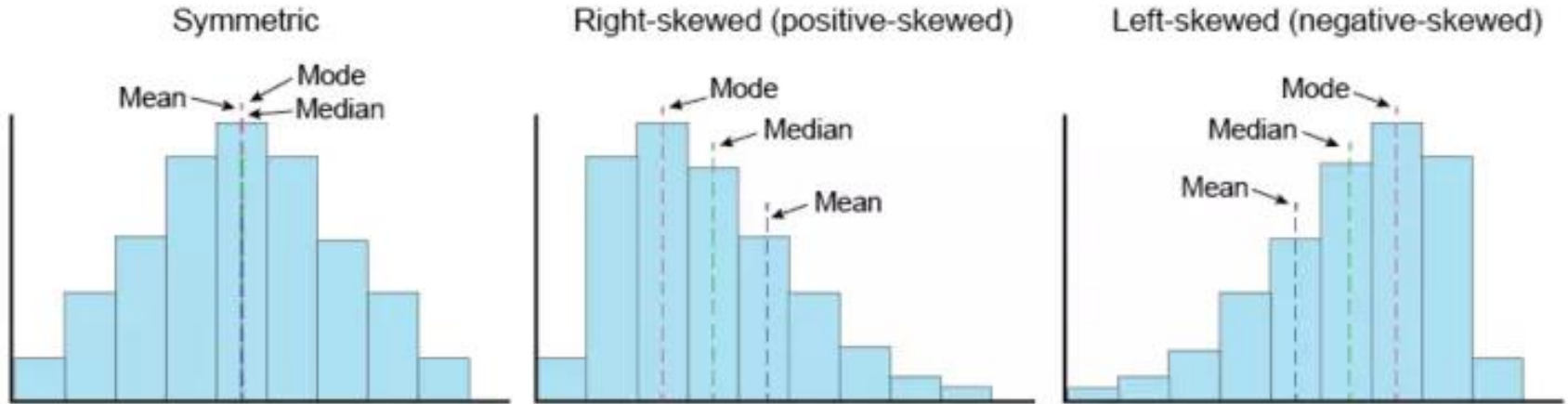


Left-Skewed (Negative Skewness)



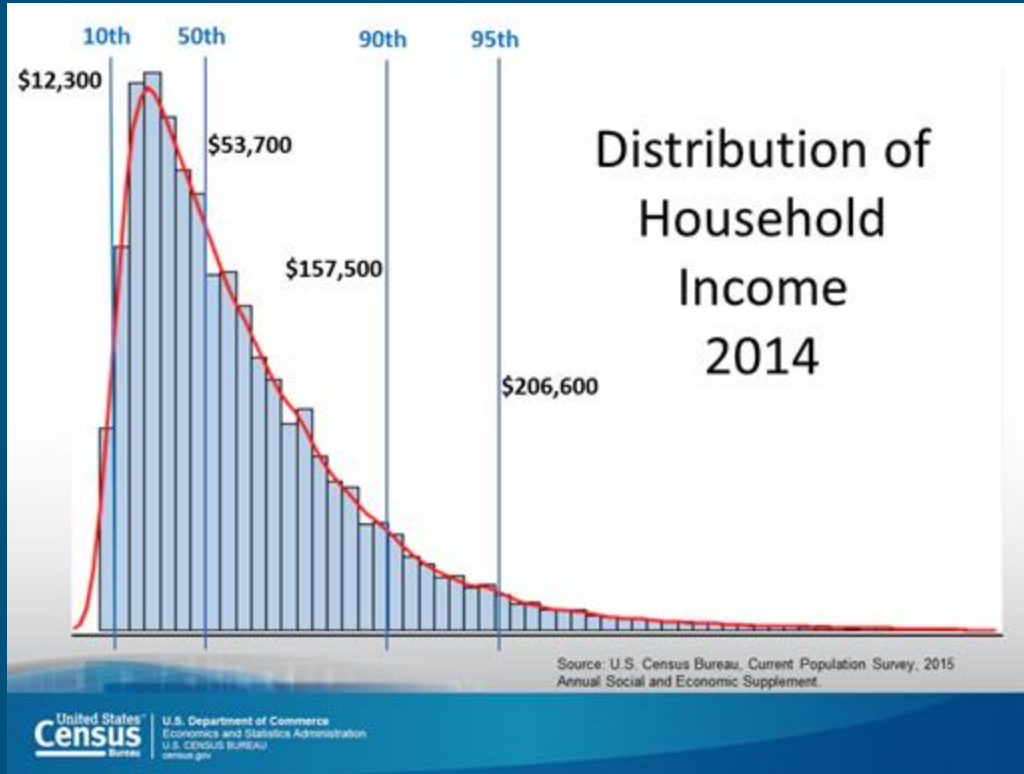
Right-Skewed (Positive Skewness)

# Histograms: Skew





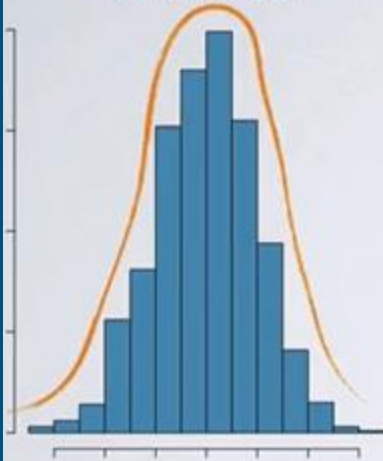
# Histograms: Skew



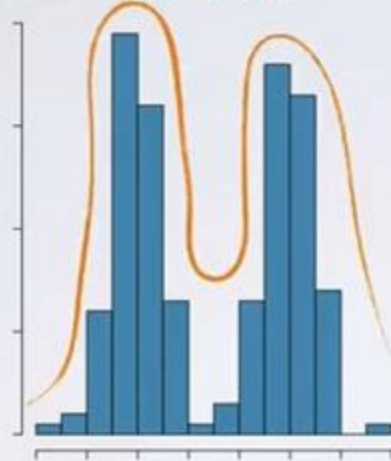
# Histograms: Modality

modality

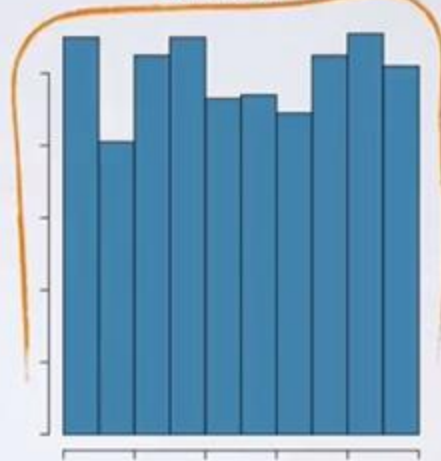
unimodal



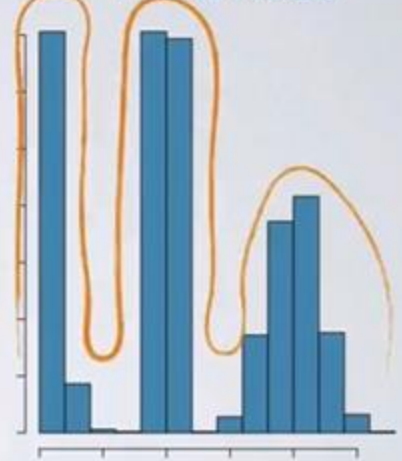
bimodal



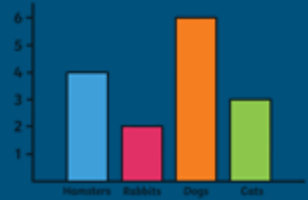
uniform



multimodal



# Visualization



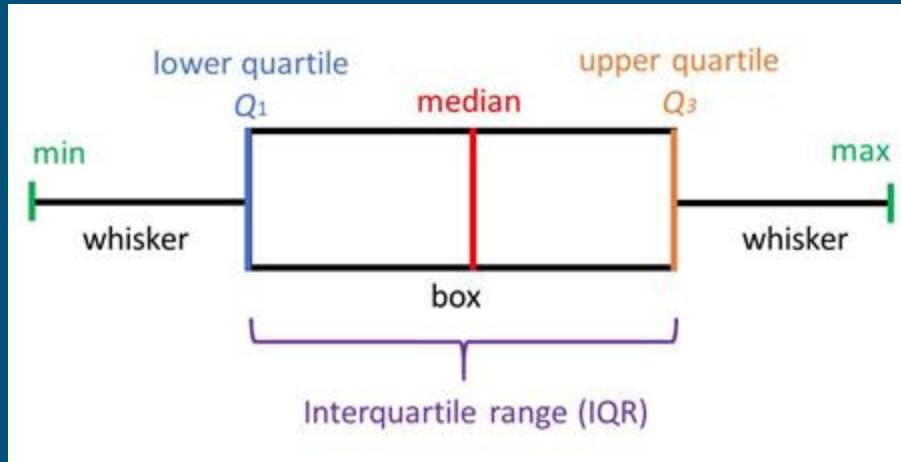
Dot plots

Frequency tables

Stem-and-leaf plots

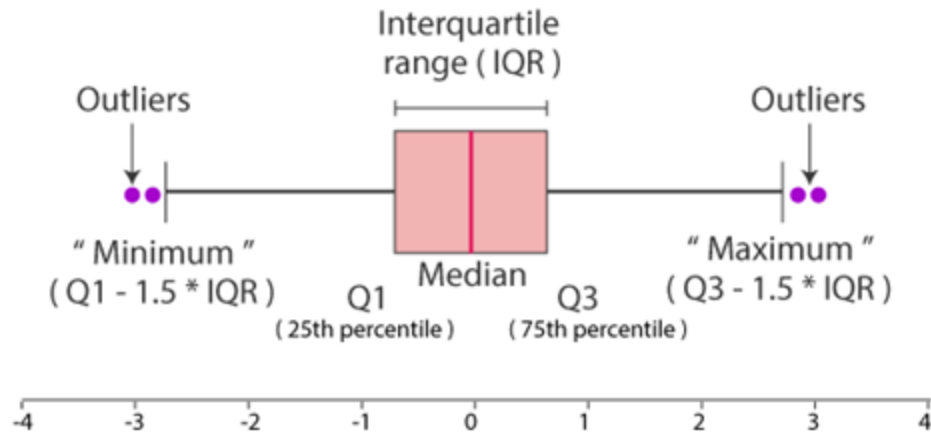
Histograms

Box plots



# Box plots

Visualization of the 5-number summary, mostly



Different parts of boxplot

# Outliers: Box plot

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$IQR \times 1.5$  = length of whiskers

Anything outside of whiskers is an outlier

Outliers are considered statistically “extreme”

# Outliers: Robust or no?

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If a measure is *robust*, it means that it is not greatly affected by outliers

- Robust: median, IQR
- Not robust: mean, standard deviation

For symmetric data sets with no large outliers, better to use mean and SD

For skewed data sets or those with outliers, use median and IQR