

DATA ANALYSIS

Week 4: Sampling

logistics: midterm 1

- computational grades will be made available **latest by tomorrow**
- review my comments in rubric
- no points were taken off for some consistent errors but I have left comments for these
 - Part 1, Q3d (ii)
 - Part 3, Q7
- you will also see a FINAL midterm grade (out of 15; conceptual + computational)
- please come see me if you have questions!
 - Monday: available from **1-3 pm (Kanbar 217)**

today's agenda



probability and inference



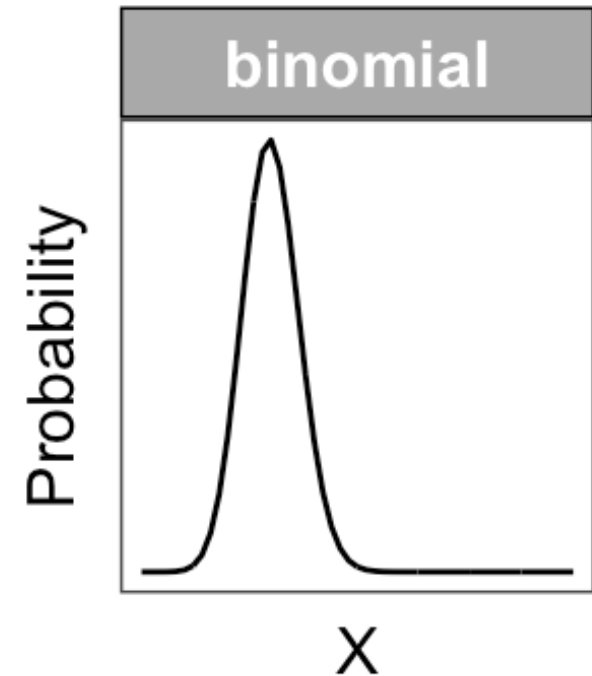
sampling



class survey

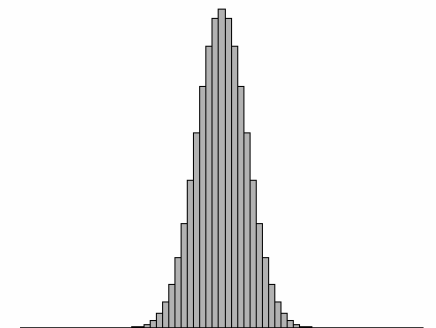
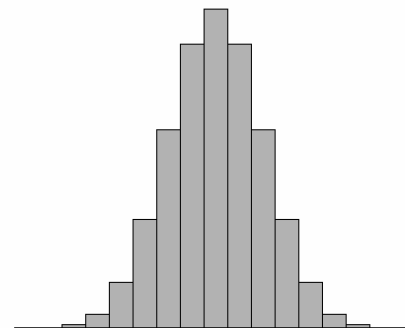
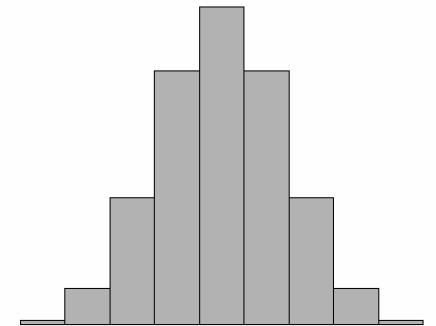
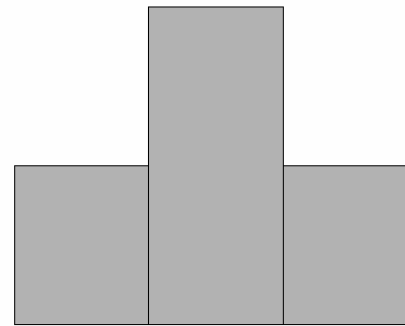
recap: binomial distribution

- data can only take two possible values (bi = two, nomial = names)
- a sequence of “bernoulli trials” (with only 2 possible outcomes)
- question of interest: how often does an outcome (A or B) occur in a sample of observations?
 $p = p(A)$ and $q = p(B)$
 $p + q = 1$ i.e., $q = 1 - p(A)$ and $p = 1 - p(B)$
- n : number of observations/individuals in the sample
- X : number of times that A occurs in the sample
 - X ranges between 0 and n
- the binomial distribution shows the probability associated with each X value from $X=0$ to $X=n$



increasing n...

- play with the [coin toss simulator](#)
 - increase number of coin tosses (n)
 - simulate flips!
- as the number of coin tosses (n) increases, the distribution starts to resemble a normal distribution!
- rule of thumb: when pn and $qn \geq 10$, the binomial distribution approximates the normal distribution
 - mean: $\mu = pn$
 - standard deviation: $\sigma = \sqrt{npq}$
 - z-score: $z = \frac{X - \mu}{\sigma} = \frac{X - pn}{\sqrt{npq}}$



example 1

- using a balanced coin, what is the probability of obtaining more than 30 heads in 50 tosses?
- balanced coin, i.e., $p = p(\text{head}) = 0.5$, $q = p(\text{tail}) = 0.5$
- $n = 50$, $X = 30$
- $\mu = pn = 0.5(50) = 25$, $qn = 0.5(50) = 25$
- pn and $qn \geq 10$ so we can proceed with normal approximation
- $\sigma = \sqrt{npq} = \sqrt{50(0.5)(0.5)} = 3.54$
- $z = \frac{X - \mu}{\sigma} = \frac{30 - 25}{3.54} = 1.18$
- look up probability in [visual calculator](#)
- $p(X > 30) = \text{less than } .119$
- note: textbook uses real limits (i.e., 30.5 here, but for simplicity stick to actual number and reframe your answer to more/less than probability obtained)



example 2

- a friend bets you that he can draw a king more than 8 times in 20 draws (with replacement) of a fair deck of cards, and he does it. Is this a likely outcome, or should you conclude that the deck is not “fair”?



example 2

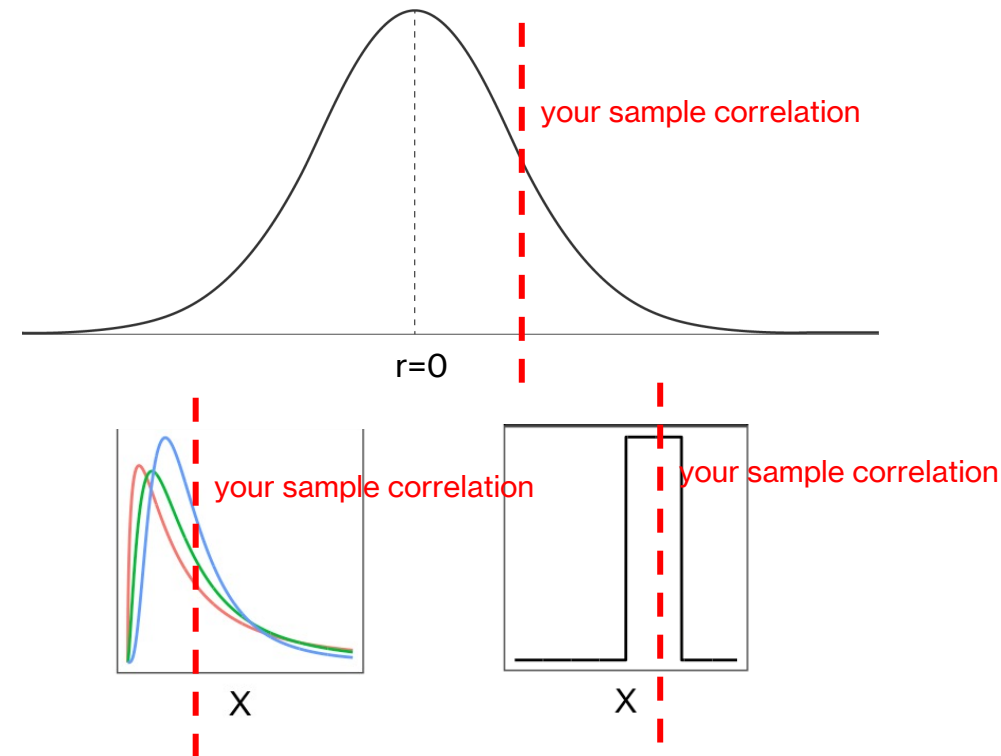
- $p(\text{king}) = \frac{4}{52} = .077, q = 1 - p = .923$
- $n = 20$ (draws), $X = 8$ (kings)
- $\mu = pn = 0.077(20) = 1.54$
- $qn = 0.923(20) = 18.46$
- $\sigma = \sqrt{npq} = \sqrt{20(0.077)(0.923)} = 1.19$
- $z = \frac{X - \mu}{\sigma} = \frac{8 - 1.54}{1.19} = 5.42$
- look up probability in [visual calculator](#)
- $p(X > 8) \approx 0!!$



three outstanding questions

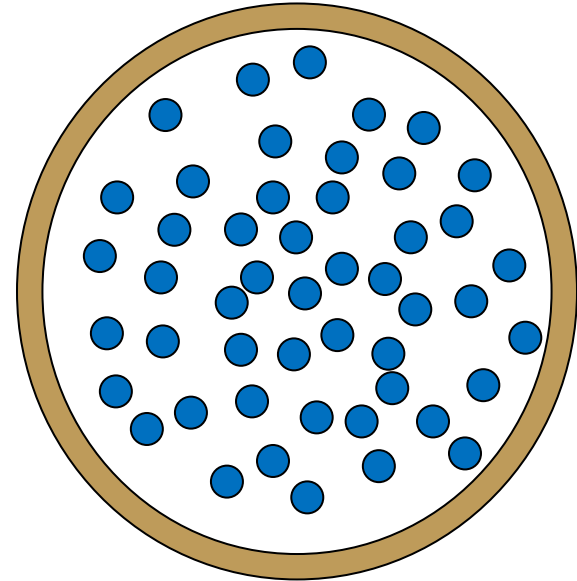
- **question 1:** how do I calculate probabilities if I don't have access to ALL the scores?
- **question 2:** how do we know what the **distribution of the null hypothesis** looks like? **If we don't know the form of the distribution, we cannot calculate probabilities**
- **question 3:** how do we know whether the probability we obtained, i.e., $P(\text{data} \mid \text{null hypothesis})$ is **small enough**?

ALL sample correlations with sample size n when there is no meaningful relationship between height and weight in the population



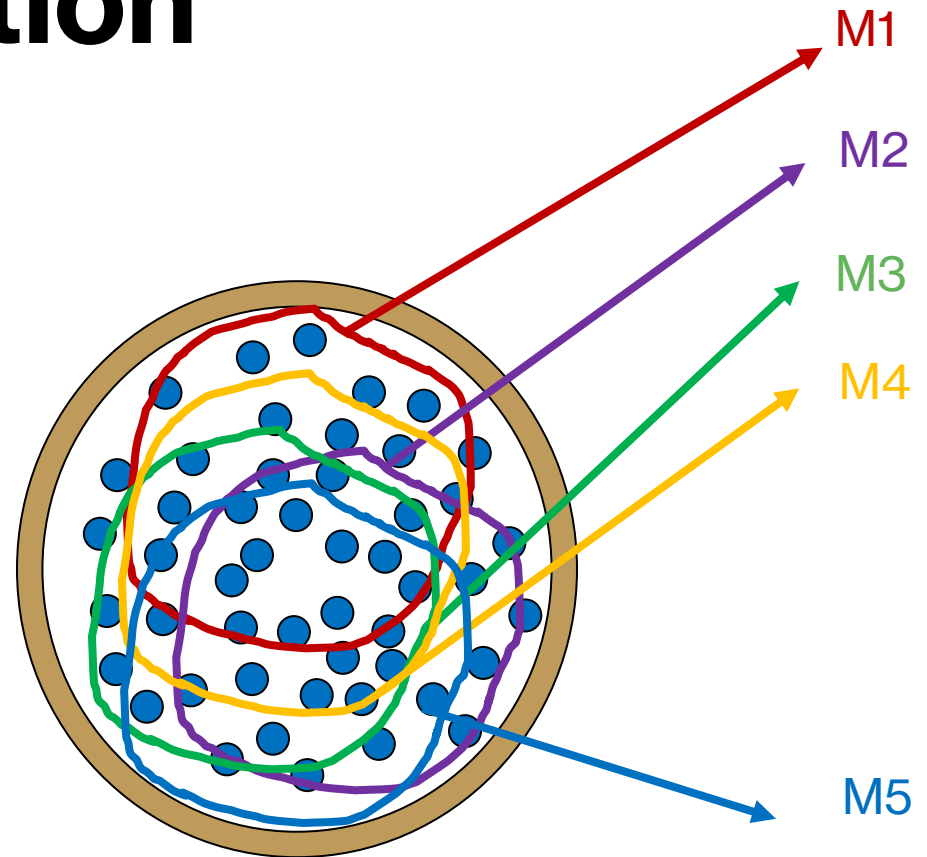
sampling from a population

- **sampling error**: the discrepancy between the sample statistic and the true population parameter it is estimating
- each time we sample, we compute some type of statistic (e.g., mean, correlation, etc.)
- **sampling distribution**: distribution of all possible values of the **statistic** obtained from multiple samples of a given size
- **distribution of sample means** contains all sample means of a size n that can be obtained from a population



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what does the sampling
distribution of means
look like??

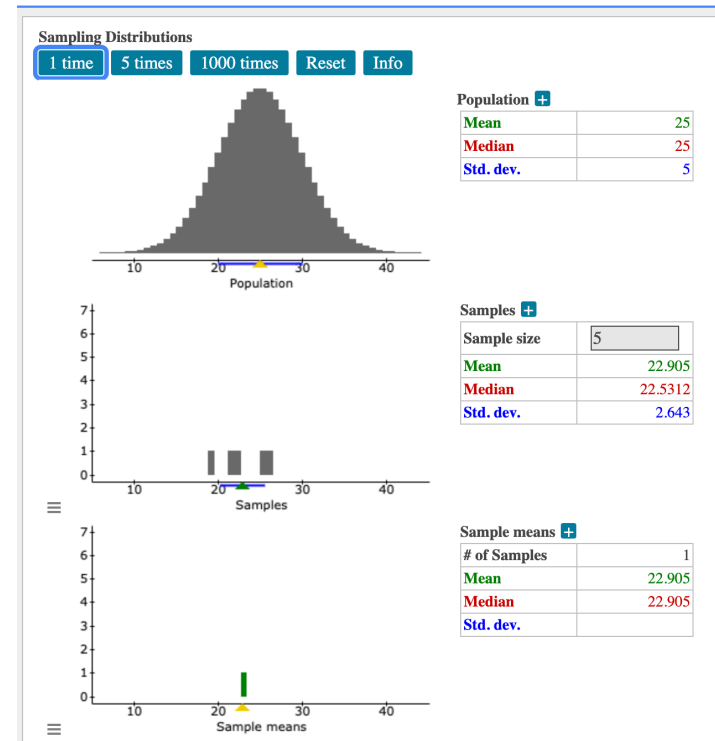
sampling distribution

- [simulator](#)
- change the distribution to bell-shaped and make sure the first statistic is the “mean” and the second statistic is “none”
- start with a single sample of size 5 and play it 1 time vs. 5 times vs. 1000 times
- explore what the three graphs are showing

Distribution
Bell shaped

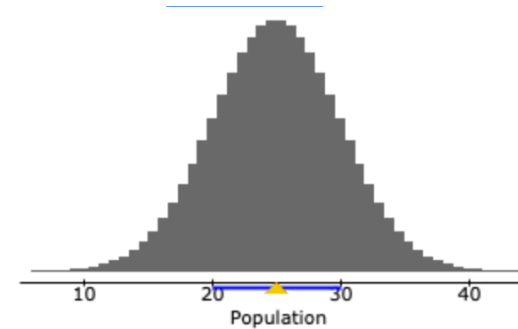
First statistic
Mean

Second statistic
None



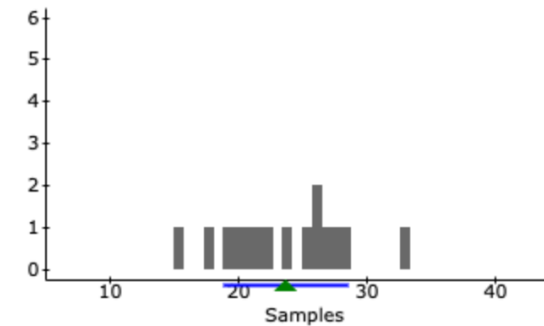
sampling distribution

- three distributions
 - the population distribution
 - the sample distribution
 - the **sampling** distribution (of all means)
- mean of sample means = population mean (unbiased estimator!)
- the sampling distribution of means approximates the normal distribution as n (sample size) increases



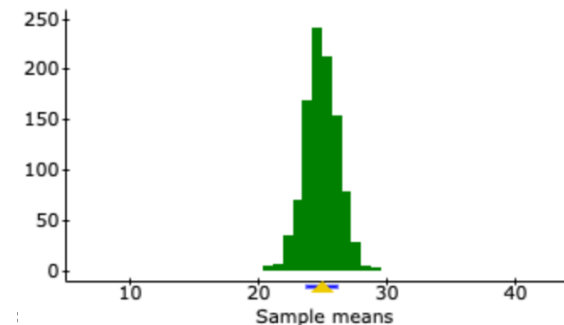
Population +

Mean	25
Median	25
Std. dev.	5



Samples +

Sample size	15
Mean	23.68
Median	24.1384
Std. dev.	4.8149



Sample means +

# of Samples	1000
Mean	24.9524
Median	24.9226
Std. dev.	1.2832

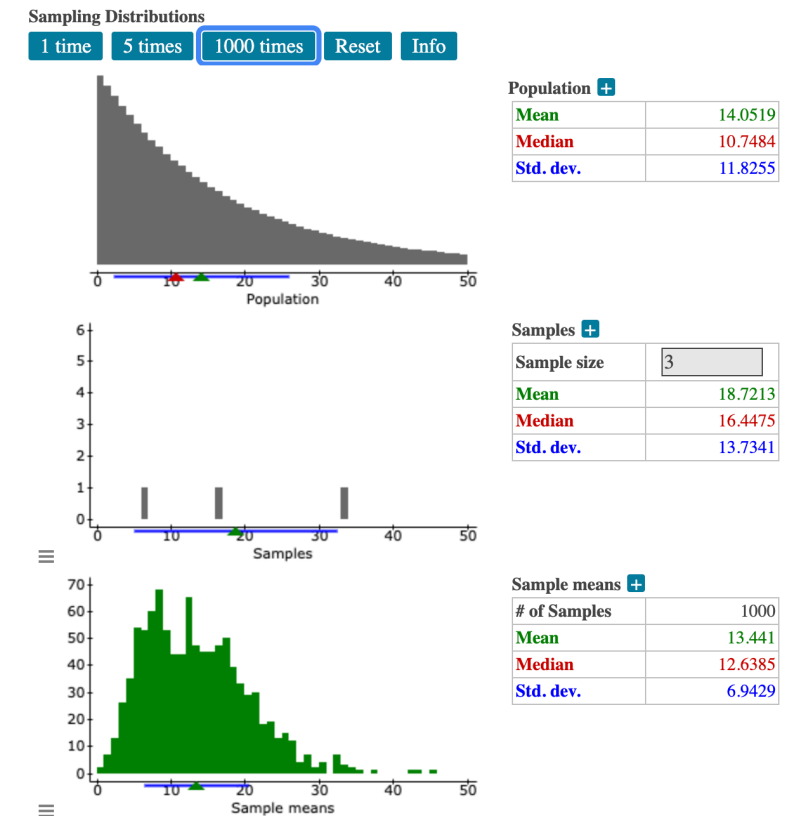
from all samples to few samples

- in practice, we cannot compute all possible samples of size n
- the **central limit theorem** states that for **any** population with mean μ and standard deviation σ , the **distribution of sample means** for sample size n will have:
 - a mean of $\mu_M = \mu$ = expected value of M
 - a standard deviation of $\sigma_M = \frac{\sigma}{\sqrt{n}}$ = standard error of the mean or M
 - will approach a normal distribution as n approaches infinity
 - distribution of sample means will be normally distributed **even if the population was not normally distributed (if n is large enough!)**
 - typically n (number of scores in a sample) around 30 yields a reasonably normal distribution

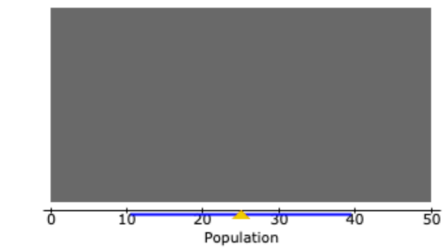
any distribution?

- [simulator](#)
- change the population distribution to any non-normal distribution
- make sure the first statistic is the “mean” and the second statistic is “none”
- explore what the sampling distribution looks like for small and large samples

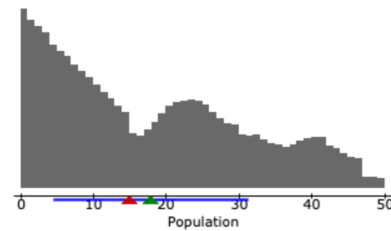
Distribution: Right skewed
First statistic: Mean
Second statistic: None



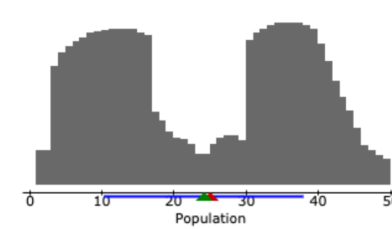
distribution of sample means



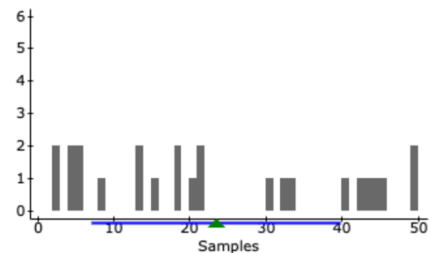
Population +	
Mean	25
Median	25
Std. dev.	14.4309



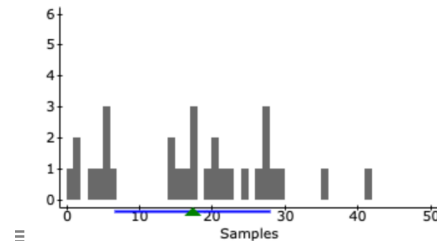
Population +	
Mean	17.8409
Median	15.0101
Std. dev.	13.3769



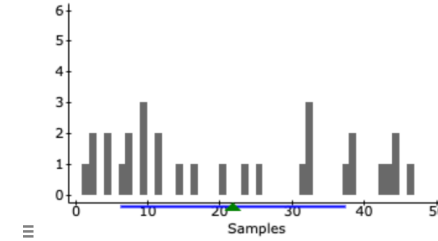
Population +	
Mean	24.157
Median	25.0403
Std. dev.	13.7207



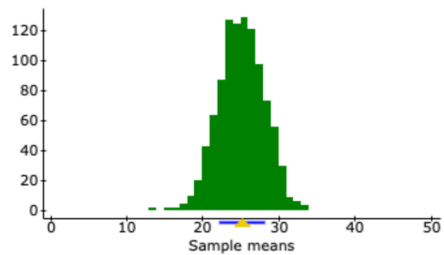
Samples +	
Sample size	25
Mean	23.5236
Median	20.6888
Std. dev.	16.3464



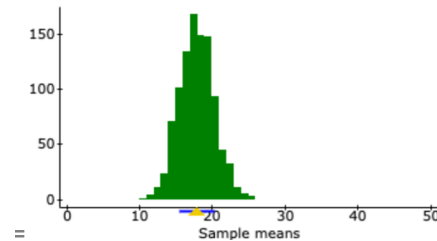
Samples +	
Sample size	30
Mean	17.3445
Median	17.694
Std. dev.	10.6863



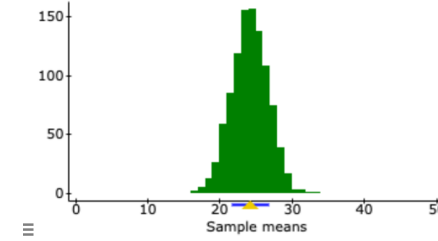
Samples +	
Sample size	30
Mean	21.8028
Median	18.6057
Std. dev.	15.5371



Sample means +	
# of Samples	1000
Mean	25.1766
Median	25.172
Std. dev.	2.929



Sample means +	
# of Samples	1000
Mean	17.9594
Median	17.8971
Std. dev.	2.3847

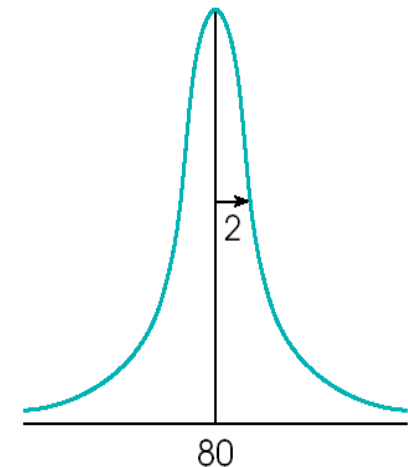
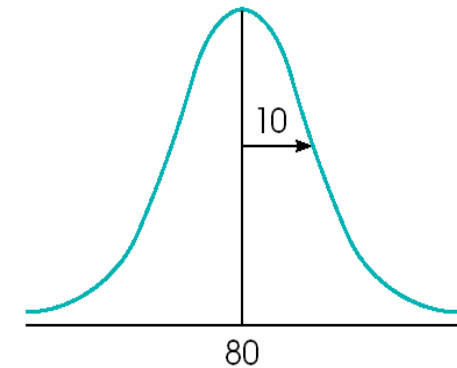
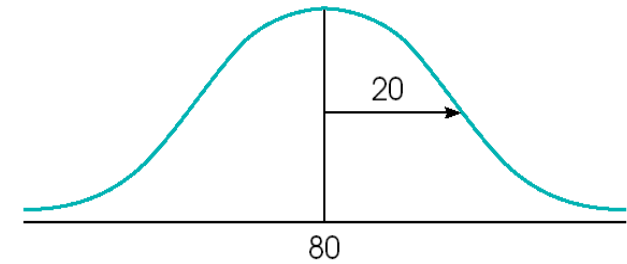


Sample means +	
# of Samples	1000
Mean	24.2364
Median	24.2623
Std. dev.	2.5189

for a large n, distribution of sample means will be normally distributed even if the population was not normally distributed!

standard error of the mean

- standard error of the mean: $\sigma_M = \sqrt{\frac{\sigma^2}{n}} = \frac{\sigma}{\sqrt{n}}$
- how different a mean from one sample could be from another on “average”
- also measures reliability: how well an individual sample’s mean represents the entire distribution of sample means
- **law of large numbers**: the larger the sample size (n), the more likely that the sample mean is closer to the population mean, and smaller the σ_M
- insight: we cannot control the population standard deviation but we can control the sample size!
- if we want our standard error of the mean to be low, we can use larger samples



example

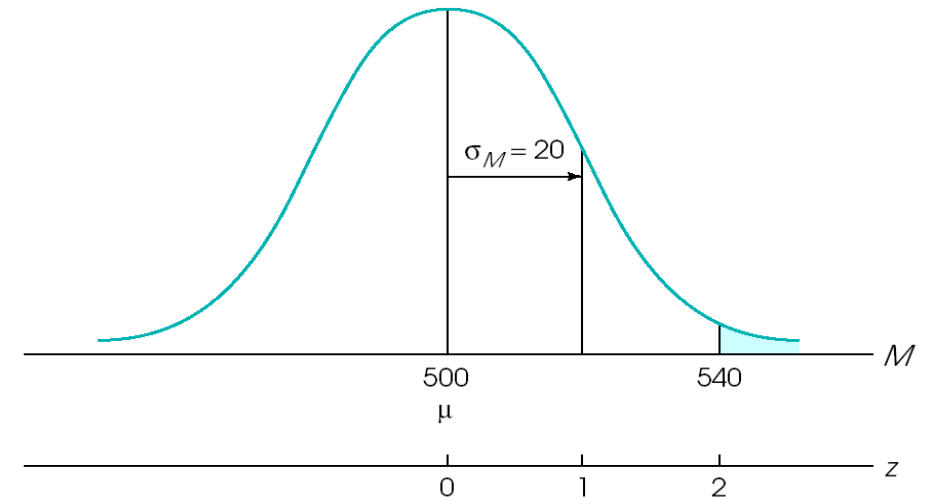
- SAT-scores population ($\mu = 500$, $\sigma = 100$). If you take a random sample of $n = 16$ students, what is the **probability that the sample mean will be greater than $M = 540$** ?
- we are talking about the sample mean, not an individual score anymore! i.e., we use the distribution of sample means (i.e., the sampling distribution) which approaches the normal distribution for large n
- represent the problem graphically
- calculate σ_M and z

$$\sigma_M = \frac{\sigma}{\sqrt{n}} = \frac{100}{\sqrt{16}} = \frac{100}{4} = 25$$

$$z = \frac{M - \mu}{\sigma_M} = \frac{540 - 500}{25} = 1.6$$

- [visual calculator](#)
- $p(M > 540) = \text{less than } .0548$

distribution of sample means



activity

- Jumbo shrimp are those that require 10–15 shrimp to make a pound. Suppose that the number of jumbo shrimp in a 1-pound bag averages $\mu = 12.5$ with a standard deviation of $\sigma = 1$, and forms a normal distribution. What is the probability of randomly picking a sample of $n = 25$ 1-pound bags that average more than $M = 13$ shrimp per bag?

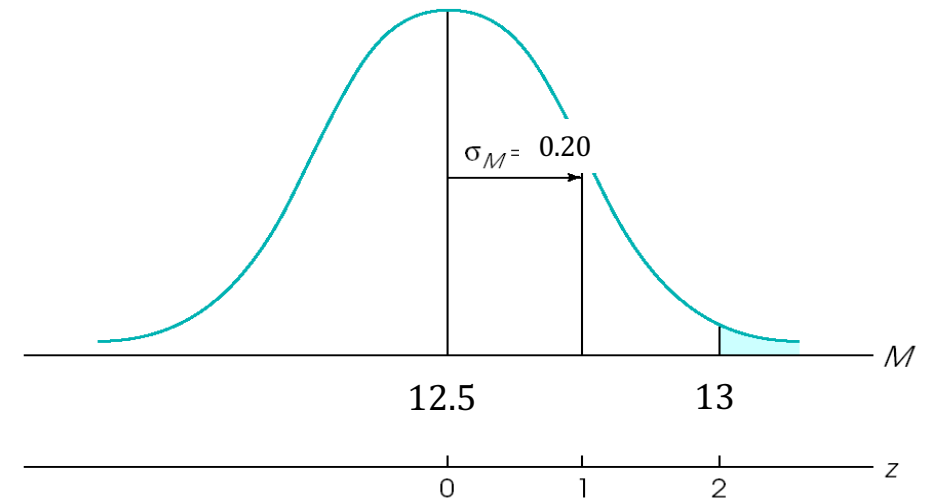
activity

- $n = 25, \mu = 12.5, \sigma = 1$
- represent the problem graphically
- calculate σ_M and z

$$\sigma_M = \frac{\sigma}{\sqrt{n}} = \frac{1}{\sqrt{25}} = \frac{1}{5} = 0.20$$

$$z = \frac{M - \mu}{\sigma_M} = \frac{13 - 12.5}{0.20} = \frac{0.5}{0.20} = 2.5$$

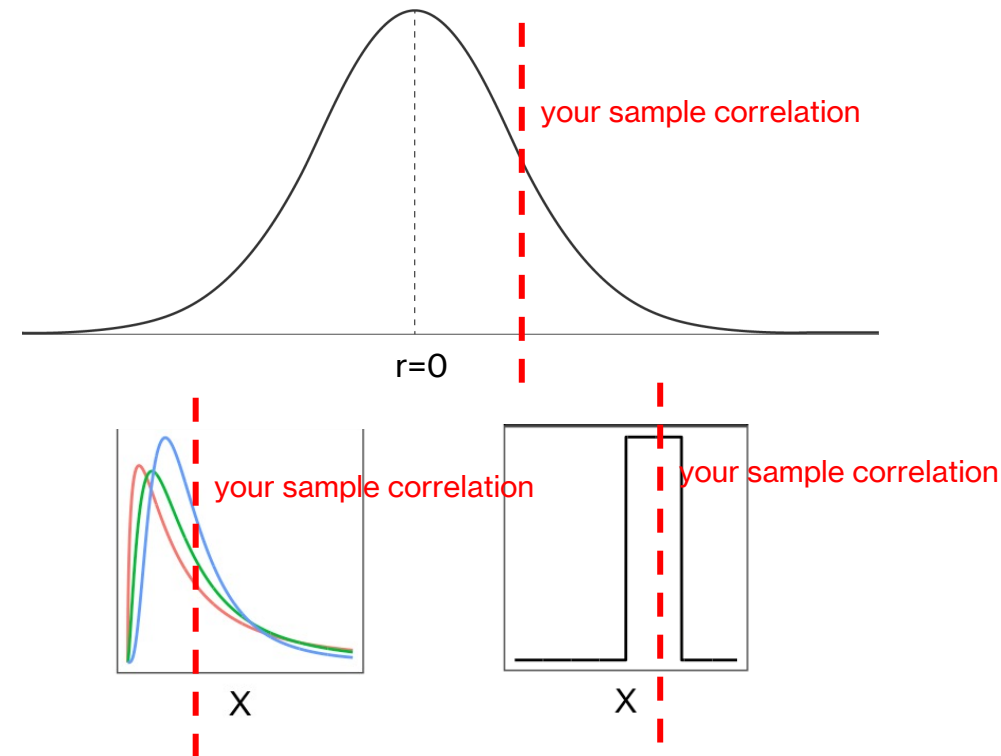
- look up probability in [visual calculator](#)
- $p(M > 13) = \text{less than } .0062$



three outstanding questions

- **question 1:** how do I calculate probabilities if I don't have access to ALL the scores?
- **question 2:** how do we know what the **distribution of the null hypothesis** looks like? **If we don't know the form of the distribution, we cannot calculate probabilities**
- **question 3:** how do we know whether the probability we obtained, i.e., $P(\text{data} \mid \text{null hypothesis})$ is **small enough**?

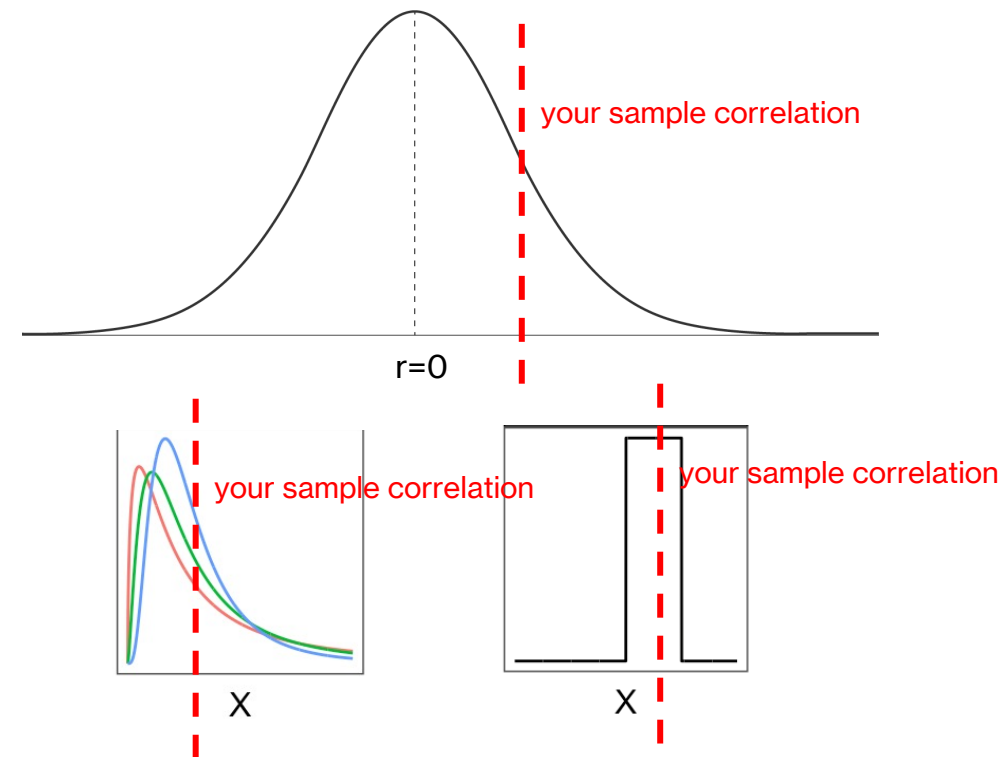
ALL sample correlations with sample size n when there is no meaningful relationship between height and weight in the population



outstanding question #1

- **question 1:** how do I calculate probabilities if I don't have access to ALL the scores?
- if I know that the distribution of the sample statistic is normal, or approaches normal, then I can calculate probabilities!

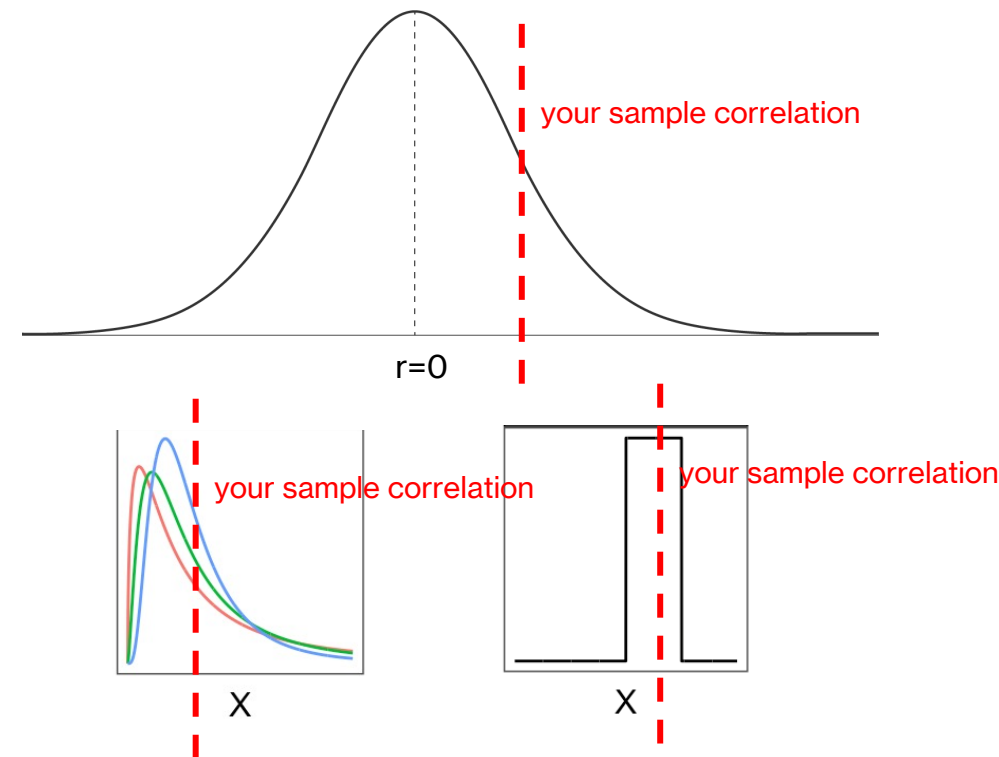
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outstanding question #2

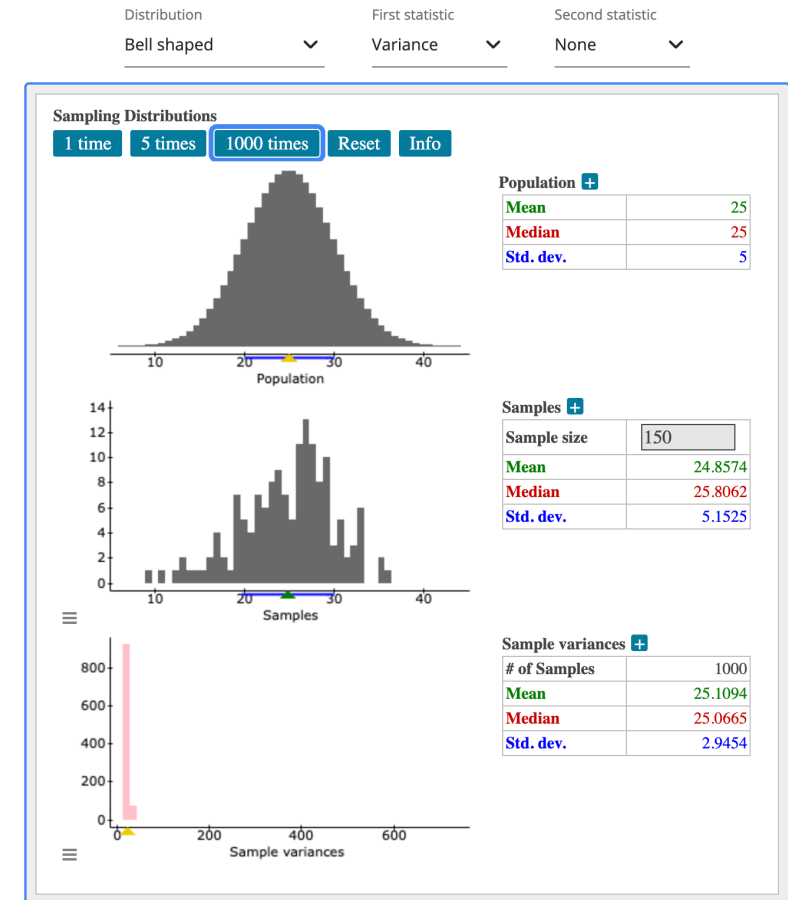
- **question 2:** how do we know what the **distribution of the null hypothesis** looks like? **If we don't know the form of the distribution, we cannot calculate probabilities**
- the central limit theorem states that for **any** population, the distribution of **sample means** will be normal for large sample ($n > 30$)
- **caveat: CLT only applies to sample means, NOT other sample statistics!**

ALL sample correlations with sample size n when there is no meaningful relationship between height and weight in the population



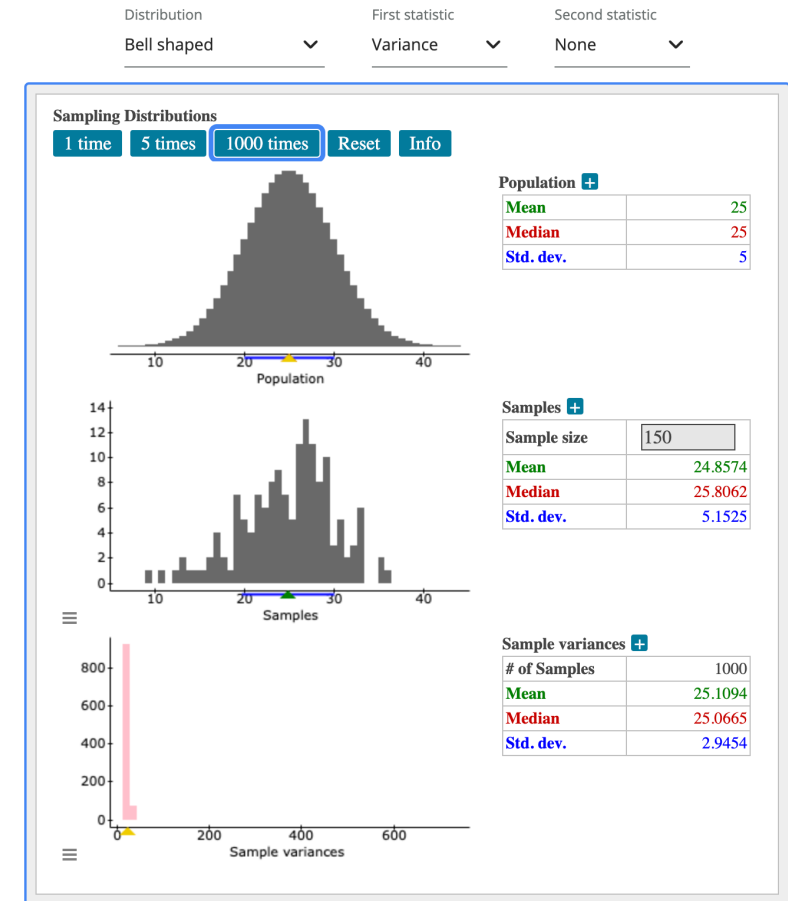
outstanding question #2

- [simulator](#)
- change the distribution to bell-shaped and make sure the first statistic is the “**variance**” and the second statistic is “none”
- start with a single sample of size 5 and play it 1 time vs. 5 times vs. 1000 times
- explore what the three graphs are showing
- the sampling distribution of variances is NOT normally distributed
- sampling distribution of several other statistics (e.g., correlation) may also NOT be normally distributed!



outstanding question #2

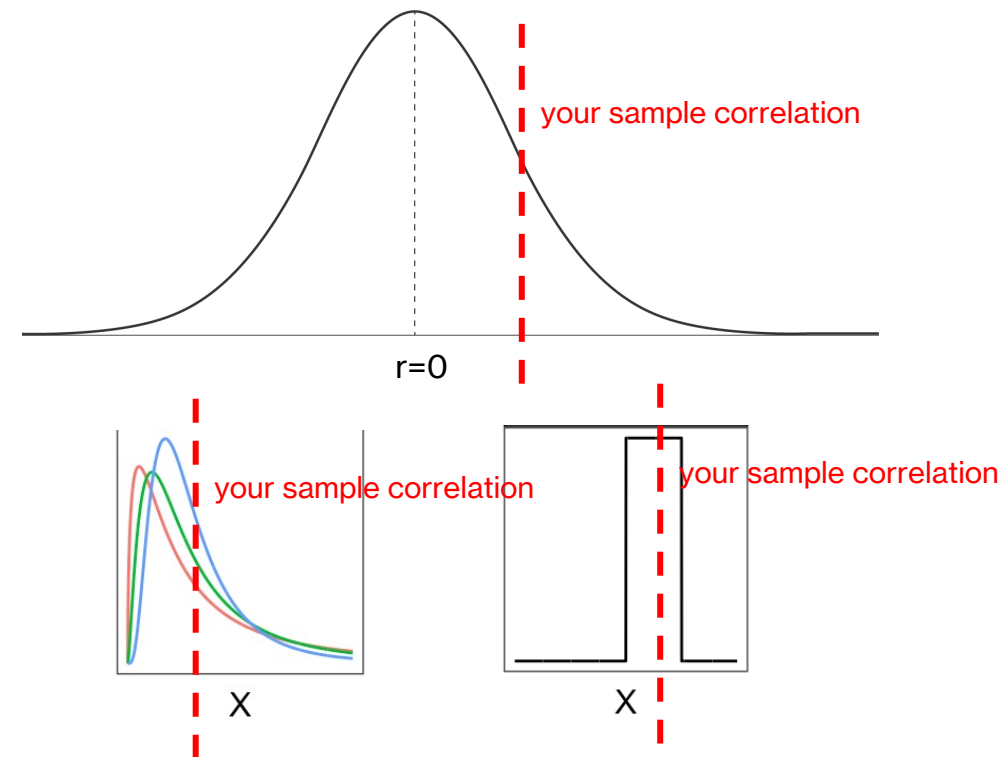
- **question 2:** how do we know what the distribution of the null hypothesis looks like? If we don't know the form of the distribution, we cannot calculate probabilities
- if we can figure out the sampling distribution of the sample statistic (e.g., means, variances, correlations, etc.), and we know the mathematical form of these distributions, we can find probabilities



outstanding question #3

- **question 3:** how do we know whether the probability we obtained, i.e., $P(\text{data} \mid \text{null hypothesis})$ is **small enough**?
- we need to set thresholds in place BEFORE we look at our data (no peeking!)
- all researchers/scientists must follow the same framework when testing hypotheses
- enter: null hypothesis significance testing (NHST)

ALL sample correlations with sample size n when there is no meaningful relationship between height and weight in the population



some changes

- problem sets now due Tuesday night
 - PS4 is due March 11
 - PS4 revision is due March 27
- quizzes will still be due Monday night
- pace will be slower
- more practice problems using Sheets (but, we have limited time)
- PLEASE watch the videos (when they are listed on the website!)
- rethinking office hour times (TBD)
- thanks for your feedback!

next time

- **before** class
 - *prep*: textbook readings
 - *try*: week 6 quiz
 - *apply*: PS4 problems (chapters 6 and 7)
 - *apply*: optional meme
- **during** class
 - hypothesis testing