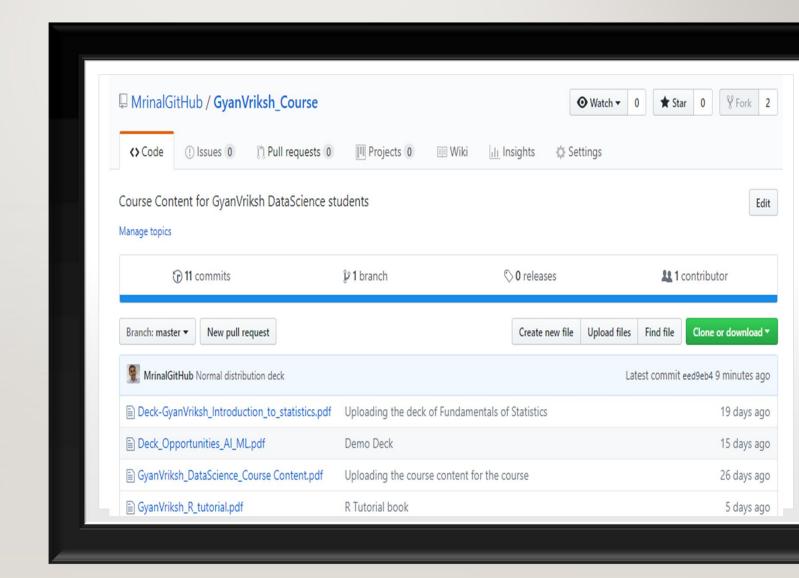
MPLING STRIBUTIONS ND THE CENTRAL 11T THEOREM



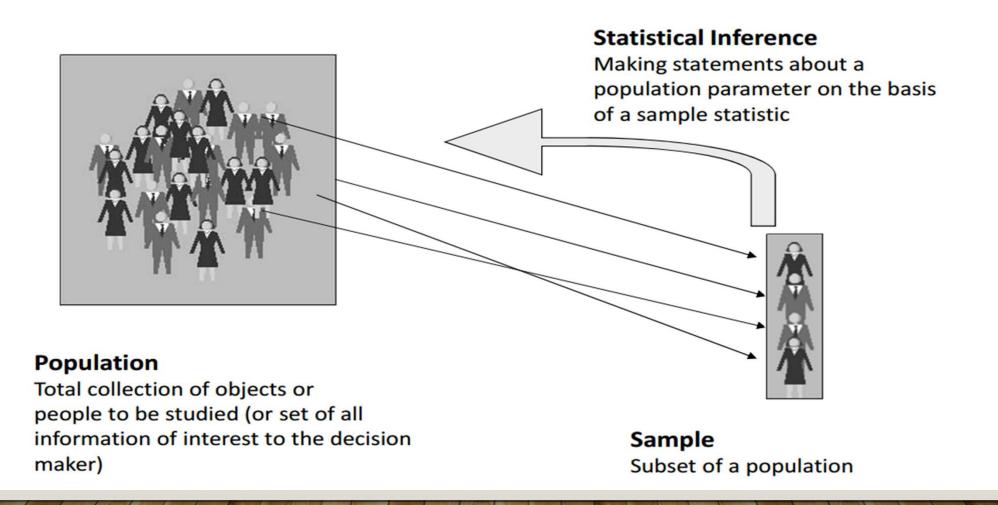


Learning objectives

- What is statistical inference?
- How to (and how not to) choose a sample?
- What are sample statistics and their properties?
- What is the central limit theorem and how is it useful?



Statistical Inference





Typical Pitfalls in Sampling

- Collecting data only from volunteers (voluntary response sample)
 - e.g. online reviews (yelp.com, maps.google.com, tripadvisor.com)
- Picking easily available respondents (convenience sample)
 - e.g. choosing to survey in In-Orbit mall
- A high rate of non-response (more than 70%)
 - e.g. CEO / CIO surveys on some industry trends



Sample statistics and population parameters

- A sample statistic is a characteristic of the sample
- Some sample statistics might be used as a point estimate for a population parameter
- We use different notations to distinguish between the two groups of numbers





Population Parameter		Sample Statistic
μ	Mean	\bar{x}
σ^2	Variance	s ²
π	Proportion	р



Selecting a Simple Random Sample (SRS)

- Unbiased: Each unit has equal chance of being chosen in the sample
- Independent: Selection of one unit has no influence on selection of other units
- SRS is a gold standard against which all other samples are measured



Selecting the Sampling Frame

- Sampling frame is simply a list of items from which to draw a sample
- Does the sampling frame represent the population?
 - e.g. Literary Digest vs. George Gallup polls
- The available list may differ from desired list
 - e.g. we don't have list of customers who did not buy from a store
- Sometimes, no comprehensive sampling frame exists
 - e.g. when forecasting for the future. Thus a comprehensive list of acceptances of credit card offers does not exist yet



Example

What is the average work experience of all participants of the BA course?

	Sample 1	Sample 2
1		
2		
3		
4		
5		
6		
7		
8		
9		
10		
Sample Mean		



Central Limit Theorem (CLT) & the distribution of the sample mean

- The distribution of the sample mean
 - will be normal when the distribution of data in the population is normal
 - will be approximately normal even if the distribution of data in the population is not normal, if the sample size is "fairly large"
- Mean $(\bar{X}) = \mu$ (the same as the population mean of the raw data)
- Standard deviation (X)= $\frac{\sigma}{\sqrt{n}}$, where σ is the population standard deviation and n is the sample size
 - This is referred to as Standard Error of the Mean

Activity: http://www.socr.ucla.edu/htmls/SOCR Experiments.html



CLT is Valid When...

- Each data point in the sample is independent of the other
- The sample size is large enough

 $J_{\mathcal{B}}$

Suppose salaries at a very large corporation have a mean of \$62,000 and a standard deviation of \$32,000.

If a single employee is randomly selected, what is the probability their salary exceeds \$66,000?

 $J_{\mathcal{B}}$

Suppose salaries at a very large corporation have a mean of \$62,000 and a standard deviation of \$32,000.

If 100 employees are randomly selected, what is the probability their average salary exceeds \$66,000?



Example and Resource material:

T-table sample link: http://www.itl.nist.gov/div898/handbook/eda/section3/eda3672.htm

Refer page 24:

https://www.saylor.org/site/wp-content/uploads/2012/03/Introductory-Business-Statistics.pdf



nmary of Session

atistical inference is the process of making probabilistic inferences about population rameters based on sample statistics

mple random sample is the gold standard and it requires a sampling frame that present the population and a randomization device

mple statistics are random variables because they vary across samples drawn from the me population. They can be used as point estimates of the population parameter.

intral limit theorem states that, no matter what the population distribution is, the mple mean ($\overline{\chi}$) is normally distributed with mean (μ) and standard error $\left(\frac{\sigma}{\sqrt{n}}\right)$, proximately