

## A Statistics Summary-sheet

	<u>Sampling Conditions</u>	<u>Confidence Interval</u>	<u>Test Statistic</u>
<b>Yes</b>	$\sigma^2$ is known $\Rightarrow \bar{X} \sim N(\mu, \sigma^2/n)$	$\bar{X} \pm Z_{\alpha/2} \left( \frac{\sigma}{\sqrt{n}} \right)$	$Z = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}}$
	$\sigma^2$ is unknown $\Rightarrow \bar{X} \sim N(\mu, \sigma^2/n)$	$\bar{X} \pm Z_{\alpha/2} \left( \frac{s}{\sqrt{n}} \right)$	$Z = \frac{\bar{X} - \mu_0}{s / \sqrt{n}}$
<b>Is n is large, say over 30?</b>			
<b>No</b>		$\bar{p} \pm Z_{\alpha/2} \left( \sqrt{\frac{\bar{p}(1-\bar{p})}{n}} \right)$	$Z = \frac{\bar{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$
	$X \sim N(\mu, \sigma^2)$ and $\sigma^2$ is known $\Rightarrow \bar{X} \sim N(\mu, \sigma^2/n)$	$\bar{X} \pm Z_{\alpha/2} \left( \frac{\sigma}{\sqrt{n}} \right)$	$Z = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}}$
	$X \sim N(\mu, \sigma^2)$ and $\sigma^2$ is unknown $\Rightarrow \bar{X} \sim t_{n-1}(\mu, \sigma^2/n)$	$\bar{X} \pm t_{n-1, \alpha/2} \left( \frac{s}{\sqrt{n}} \right)$	$t_{n-1} = \frac{\bar{X} - \mu_0}{s / \sqrt{n}}$

**If n is not large, say over 30 and X is not  $\sim N(\mu, \sigma^2)$ , cannot proceed with parametric statistics.**

## *Formulas, Distributions, and Concepts*

### Counting and Probabilities

$${}_n P_x = \frac{n!}{(n-x)!} \text{ Permutations}$$

$${}_n C_x = \frac{n!}{x!(n-x)!} \text{ Combinations}$$

$$P(A | B) = \frac{P(A \cap B)}{P(B)} \text{ Conditional Probability}$$

$$P(A \cap B) = P(A | B)P(B) \text{ Probability of an Intersection}$$

### Discrete Probability Distributions

$$P_x(x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x} \text{ Binomial Probability}$$

$$P_x(x) = \frac{e^{-\mu} \mu^x}{x!} \text{ Poisson Probability}$$

### Continuous Probability Distributions

**Random Variable** ~ Distribution (mean, variance)

Standard Normal       $Z \sim N(0,1)$

Normal  $X \sim N(\mu, \sigma^2)$

Binomial  $X \sim \text{Binomial}[np, np(1-p)]$

Sample Mean  $\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$

By CLT, if  $n \geq 30$ ,  $\bar{X} \sim N\left(\mu, \frac{\sigma^2}{n}\right)$

Sample Proportion  $\bar{p} \sim \left(p, \frac{p(1-p)}{n}\right)$

### **Confidence Intervals (Interval Estimation)**

$\bar{X} \pm z_{(\alpha/2)} \frac{\sigma}{\sqrt{n}}$  If population is normal and population variance is known.

$\bar{X} \pm z_{(\alpha/2)} \frac{s}{\sqrt{n}}$  If population variance is unknown and  $n \geq 30$ .

$$\bar{X} \pm t_{(n-1, \alpha/2)} \frac{s}{\sqrt{n}} \quad \text{If population is normal, population variance is unknown.}$$

$$\bar{p} \pm z_{(\alpha/2)} \sqrt{\frac{\bar{p}(1-\bar{p})}{n}} \quad \text{If } n \geq 30.$$

$$(\bar{X} - \bar{Y}) \pm z_{(\alpha/2)} \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}} \quad \text{If independent samples and either population variance known, or } n \geq 30 \text{ in which case, substitute sample variance for population variance.}$$

$$(\bar{X} - \bar{Y}) \pm t_{(n_X+n_Y-2, \alpha/2)} \sqrt{s^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)} \quad \text{where}$$

$$s^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2} \quad \text{If independent samples, population variances unknown, but statistically equal.}$$

### Estimating Sample Size

$$n = \frac{z_{\alpha/2}^2 \sigma^2}{E^2}$$

For estimation and CI for the population mean, normal population,  $\sigma^2$  known, or estimated by a pilot run. E = absolute error.

### Hypothesis Testing

1. Set up the **appropriate** null which must be in equality form, always and alternative hypotheses.
2. Define the rejection area. Take care as to whether the test is one-tailed or two-tailed. Look to the alternative hypothesis to determine this.
3. Calculate the test statistic.
4. State Decision.
5. Interpret your conclusion.

### Hypothesis Testing (test statistics and their distributions under the null)

$$\frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}} \sim z_{\alpha}$$

When population variance known, or if  $n \geq 30$ , substitute  $s$  for  $\sigma$ .

$$\frac{\bar{X} - \mu_0}{s / \sqrt{n}} \sim t_{n-1, \alpha}$$

When If population is normal, population variance unknown.

$$\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)_0}{\sqrt{s^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}} \sim t_{n_1+n_2-2, \alpha} \quad \text{where } s^2 = \frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2} \quad \text{Difference in means, independent}$$

samples, population variances unknown, but statistically almost equal.

$$\frac{\bar{p} - p_0}{\sqrt{p_0(1 - p_0)/n}} \sim Z_{\alpha} \quad \text{Population Proportion, } n \geq 30$$

## Formulas for ANOVA

$$\bar{X}_j = \frac{\sum_{i=1}^{n_j} x_{ij}}{n_j}$$

$$s_j^2 = \frac{\sum_{i=1}^{n_j} (x_{ij} - \bar{X}_j)^2}{n_j - 1}$$

$$\bar{\bar{X}} = \frac{\sum_{j=1}^k \sum_{i=1}^{n_j} x_{ij}}{n_T} \quad \text{Overall Sample Mean}$$

or if the treatment sample sizes are all equal,  $\bar{\bar{X}} = \frac{\sum_{j=1}^k \bar{X}_j}{k}$

$$MSTR = \frac{SSTR}{k-1}$$

$$SSTR = \sum_{j=1}^k n_j (\bar{X}_j - \bar{\bar{X}})^2$$

$$MSE = \frac{SSE}{n_T - k}$$

$$SSE = \sum_{j=1}^k (n_j - 1) s_j^2$$

$$F = \frac{MSTR}{MSE} \sim F_{k-1, n_T - k}$$

### **Terms and Concepts**

Central Limit Theorem: If the sample size  $n$  is large, say  $n \geq 30$  no matter what the population distribution is, the sampling distribution of the sample mean tends towards the normal as  $n$  gets large.