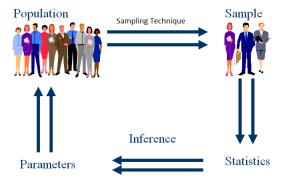
# Chapter 1

# ST 512 - Review

Readings: Chapters 1-8 as needed

# Big ideas in stats:

- Population all the values, items, or individuals of interest
- Parameter a (usually) unknown summary value about the population
- <u>Sample</u> a subset of the population we observe data on
- <u>Statistic</u> a summary value calculated from the sample observations



Examples of paramters - (true) mean  $\mu$ , (true) variance  $\sigma^2$ .

Examples of statistics - sample mean  $\bar{y}$ , sample variance  $s^2 = \frac{\sum_{i=1}^{n} (y_i - \bar{y})^2}{n-1}$ Inference - Making mathematically sound claims about the poulation using sample data.

# Scales (Types) of Data:

• Qualitative or Categorical - A variable that is described by attributes or labels Subscales:

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Nominal - categories have no ordering (Male, Female)
Ordinal - can order categories (Lickert scale data)
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• Quantitative - A variable that is described by numerical measurements where arithmetic can be performed

Subscales:

Discrete - finite or countable finite number of values (# of flowers on a plant, 0, 1, 2, ...) Continuous - any value in an interval is possibel (Temperature,  $(-459.67 \deg F, \infty)$ 

### Random Variables and Things of Interest:

• Random Variable (RV) - Function that takes in outcomes from an experiment and outputs real numbers, or a numeric outcome to a random process

#### Things of interest

- <u>Distribution</u> pattern and frequency of observable values
   For continuous RVs, visualized with a smooth curve.
- Mean/Median measures of center of the distribution

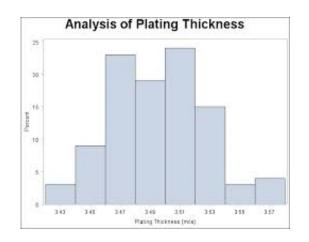
Focus on mean: true mean  $\mu$ , RV sample mean  $\bar{Y}$ , observed sample mean  $\bar{y}$ 

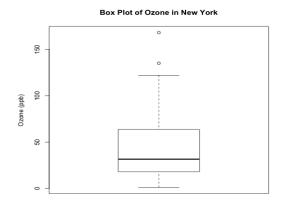
- <u>Standard Deviation, Variance, IQR, Range</u> - measures of spread for the distribution

Focus on SD and Variance: true variance  $\sigma^2$ , true SD  $\sigma$ , observed sample variance  $s^2$ , observed SD s

#### Graphical Descriptions of RV's:

- $\bullet$  <u>Histogram</u> Graphs the frequencies or relative frequencies of realizations of a RV
- Boxplot Uses the Five Number Summary to display the realizations of a RV Five number summary: min,  $Q_1$ , M,  $Q_3$ , max





Statistics are also RVs. The distribution of a statistic is called a  $\underline{\text{sampling distribution}}$  Central Limit Theorem (CLT):

If a RV Y has a (true) mean  $\mu$  and (true) variance  $\sigma^2$ , and a random sample is of size  $n \geq 30$  is taken then

$$\bar{Y} \sim N(\mu, \sigma^2/n)$$

Note: If  $Y \sim N(\mu, \sigma^2)$  then  $\bar{Y} \sim N(\mu, \sigma^2/n)$  for any n.

2 main ways to make inference about a (true) mean,  $\mu$ :

1. When the true SD,  $\sigma$ , is known we looked at the sampling distribution of the statistic

 $Z = \frac{\bar{Y} - \mu}{\sigma/n} \sim N(0, 1)$  valid if  $\bar{Y}$  has a normal distribution

Allows us to form a CI:

And a test statistic: Testing 
$$H_0: \mu = \mu_0$$

$$\bar{y} \pm z_{\alpha/2} \sigma / \sqrt{n}$$

$$z_{obs} = \frac{\bar{y} - \mu_0}{\sigma / \sqrt{n}}$$

2. When the true SD,  $\sigma$ , is unknown we looked at the sampling distribution of the statistic

 $T = \frac{\bar{Y} - \mu}{s/n} \sim t_{n-1}$  valid if  $\bar{Y}$  has a normal distribution, allow for  $n \geq 15$  or so in CLT

Allows us to form a CI:

And a test statistic: Testing 
$$H_0: \mu = \mu_0$$

$$\bar{y} \pm t_{(n-1,\alpha/2)} s / \sqrt{n}$$

$$t_{obs} = \frac{\bar{y} - \mu_0}{s/\sqrt{n}}$$

# Inference about two (true) means, $\mu_1$ and $\mu_2$ :

• From paired samples,  $x_1, x_2, ..., x_n$  and  $y_1, y_2, ..., y_n$  where difference is normally distributed

CI: 
$$(\bar{x} - \bar{y}) \pm t_{(n-1,\alpha/2)} s_{diff} / \sqrt{n}$$

HT: 
$$H_0: \mu_1 = \mu_2$$
, i.e.  $\mu_1 - \mu_2 = 0$   $t_{obs} = \frac{(\bar{x} - \bar{y}) - 0}{s_{diff}/\sqrt{n}}$ 

• Two separate samples from normal populations,  $x_1, x_2, ..., x_n$  and  $y_1, y_2, ..., y_n$ 

CI: 
$$(\bar{x} - \bar{y}) \pm t_{(\nu,\alpha/2)} \sqrt{s_X^2/n + s_Y^2/m}$$
 where  $\nu$  is an estimate of df

HT: 
$$H_0: \mu_1 = \mu_2$$
, i.e.  $\mu_1 - \mu_2 = 0$   $t_{obs} = \frac{(\bar{x} - \bar{y}) - 0}{\sqrt{s_X^2/n + s_Y^2/m}}$ 

# Extension to inference about t (true) means, $\mu_1, \mu_2, ..., \mu_t$ :

Balanced One-way ANOVA table (same number of replicates per group)

Source	DF	SS	MS	F-stat	P-value
Treatment	t-1	$n\sum_{i=1}^{t} (\bar{Y}_{i+} - \bar{Y}_{++})^2$	$\frac{SS(Trt)}{t-1}$	$\frac{MS(Trt)}{MS(E)}$	Use $F(t-1,t(n-1))$
Error	t(n-1)	$\sum_{i=1}^{t} \sum_{j=1}^{n} (Y_{ij} - \bar{Y}_{i+})^2$	$\frac{SS(E)}{t(n-1)}$		
		$ \sum_{i=1}^{t} \sum_{j=1}^{n} (Y_{ij} - \bar{Y}_{++})^2 $			

Analysis used for a completely randomized design.

P-value tests  $H_0: \mu_1 = \mu_2 = ... = \mu_t$  vs  $H_A:$  at least 1 mean differs

One Way ANOVA model:

$$Y_{ij} = \mu + \alpha_i + E_{ij}$$

where i = 1, 2, ...t and j = 1, 2, ..., n (total sample size = nt = N)

 $\mu = \text{overall mean}$ 

 $\alpha_i$  = effect from group i

 $E_{ij}$  = random error assumed to be  $iid\ N(0,\sigma^2)$ 

# For two quantitative variables measured on the same units, the linear relationship can be investigated:

Simple linear regression model  $Y_i = \beta_0 + \beta_1 x_i + E_i$  or use correlation.

### For a hypothesis test, the p-value means

probability of observign a test statistic as extreme or more extreme than the one observed, assuming the null hypothesis is true.

## For a given a null hypothesis, statistical significance implies

the observed value was unlikely to have occurred by random chance alone (assuming the null hypothesis is true).

## For an observed confidence interval (cL, cU) we can say

We are  $\_\_$ % confident the true parameter value is contained in the interval. (\*\*\*Do not say probability or chance!)

#### The idea of Confidence means

The procedure used to create the interval has a \_\_\_% probability of producing an interval that contains the parameter.

i.e. If the experiment were done repeatedly and an interval made for each sample, \_\_\_% of the intervals would contain the parameter value.