# Introduction to Statistical Significance

BUS 735: Business Decision Making and Research

# 1 Goals and Agenda

## 1.1 Goals

## Goals

Re-familiarize ourselves with basic statistics ideas: sampling distributions, hypothesis tests, p-values.

#### Learning Outcomes

- Background for learning outcomes LO1 and L02 regarding methods of statistical analysis
- LO6: Be able to use standard computer packages such as R to conduct statistical analysis

## Agenda

Learning Objective	Active Learning Activity
Re-familiarize ourselves with ba-	Lecture / Discussion
sic statistics ideas: sampling dis-	
tributions, hypothesis tests, p-	
values.	
Get comfortable with R environ-	Online Tutorial
ment and programming language	

# 2 Statistical Significance

## 2.1 Sampling Distribution

## **Probability Distribution**

**Probability distribution:** summary of all possible values a variable can take along with the probabilities in which they occur.



$$f(x|\mu,\sigma) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\left(\frac{(x-\mu)^2}{2\sigma^2}\right)}$$

Computer (R example) > pnorm(1.0) returns P(z < 1.0) = 0.8413

## **Probability Distribution**

Probability distributions are typically defined by...

- 1. Measure of center, such as the mean of the distribution
- 2. Measure of spread, such as the variance or standard deviation
- 3. Shape, eg. symmetric, bell-shaped, defined explicitly with an equation

## **Example: Estimated Income Distribution**



## Sampling distribution

• Imagine taking a sample of size 100 from a population and computing some kind of statistic.

- The statistic you compute can be anything, such as: mean, median, proportion, difference between two sample means, standard deviation, variance, or anything else you might imagine.
- Suppose you repeated this experiment over and over: take a sample of 100 and compute and record the statistic.
- A sampling distribution is the probability distribution of the statistic
- Is this the same thing as the probability distribution of the population? NO! They may coincidentally have the same shape though.

## Sampling Distribution Simulator

Sampling Distribution Simulator http://onlinestatbook.com/stat\_sim/sampling\_dist/

## **Example in R: Create Hypothetical Population**

# Generate random pop. w/ 10,000 obs from a Chi-Square dist. pop <- rchisq(10000, 1) # Compute the population mean mean(pop) ## [1] 0.9804668 # Compute the population variance var(pop) ## [1] 1.873752 # Compute the population std dev sqrt(var(pop)) ## [1] 1.368851

#### Example in R

hist(pop,col='lightblue')

## Histogram of pop



Population mean = 0.98 Population std dev = 1.369 Population is skewed to the right

## **Generate Samples**

```
# Generate one sample of size 30
sam1 <- sample(pop,30)
mean(sam1)
## [1] 0.7461987
sqrt(var(sam1))
## [1] 0.8968072</pre>
```

hist(sam1,col='lightgreen')





## **Generate Sampling Distribution**

```
# Generate 50,000 samples,
# Each 30 obs, compute each mean
lotsmeans <-
replicate(50000,
mean( sample(pop,30) ) )
# Mean of all the means
mean( lotsmeans )
## [1] 0.9824764
# Std dev of all the means
sqrt(var(lotsmeans))
## [1] 0.2498053
```

```
# Histogram of all the means
hist(lotsmeans, col='darkgreen')
```



## Purpose of a Sampling Distribution

- In reality, you only do an experiment once, so the sampling distribution is a hypothetical distribution.
- Why are we interested in this?

#### **Desirable qualities**

What are some qualities you would like to see in a sampling distribution?

- The average of the sample statistics is equal to the true population parameter.
- Want the variance of the sampling distribution to be as small as possible. Why?
- Want the *sampling distribution* to be normal, regardless of the distribution of the population.

## 2.2 Central Limit Theorem

## **Central Limit Theorem**

- Given:
  - Suppose a RV x has a distribution (it need not be normal) with mean  $\mu$  and standard deviation  $\sigma$ .
  - Suppose a sample mean  $(\bar{x})$  is computed from a sample of size n.
- Then, if n is sufficiently large, the sampling distribution of  $\bar{x}$  will have the following properties:
  - The sampling distribution of  $\bar{x}$  will be normal.

 The mean of the sampling distribution will equal the mean of the population (unbiased):

$$\mu_{\bar{x}} = \mu$$

- The standard deviation of the sampling distribution will decrease with larger sample sizes, and is given by:

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

## Central Limit Theorem: Small samples

If n is small (rule of thumb for a single variable: n < 30)

- The sample mean is still *unbiased*.
- Formula for the standard deviation of sampling distribution still valid
- Given a small sample size, standard deviation of sampling distribution may be large
- Sampling distribution will be normal *if the distribution of the population is normal*

#### Example 1

Suppose average birth weight is  $\mu = 7lbs$ , and the standard deviation is  $\sigma = 1.5lbs$ .

What is the probability that a sample of size n = 30 will have a mean of 7.5*lbs* or greater?

$$z = \frac{\bar{x} - \mu_{\bar{x}}}{\sigma_{\bar{x}}} = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$$
$$z = \frac{7.5 - 7}{1.5/\sqrt{30}} = 1.826$$

The probability the sample mean is greater than 7.5lbs is:

$$P(\bar{x} > 7.5) = P(z > 1.826) = 0.0339$$

## Example 2

Suppose average birth weight is  $\mu = 7lbs$ , and the standard deviation is  $\sigma = 1.5lbs$ .

What is the probability that a randomly selected baby will have a weight of 7.5lbs or more? What do you need to assume to answer this question? Must assume the population is normally distributed. Why?

$$z = \frac{x - \mu}{\sigma} = \frac{7.5 - 7}{1.5} = 0.33$$

The probability that a baby is greater than 7.5lbs is:

$$P(x > 7.5) = P(z > 0.33) = 0.3707$$

## Example 3

- Suppose average birth weight of all babies is  $\mu = 7lbs$ , and the standard deviation is  $\sigma = 1.5lbs$ .
- Suppose you collect a sample of 30 newborn babies whose mothers smoked during pregnancy.
- Suppose you obtained a sample mean  $\bar{x} = 6lbs$ . If you assume the mean birth weight of babies whose mothers smoked during pregnancy has the same sampling distribution as the rest of the population, what is the probability of getting a sample mean this low or lower?

## Example 3 continued

$$z = \frac{x - \mu_{\bar{x}}}{\sigma_{\bar{x}}} = \frac{x - \mu}{\sigma/\sqrt{n}}$$
$$z = \frac{6 - 7}{1.5/\sqrt{30}} = -3.65$$

The probability the sample mean is less than or equal to 6lbs is:

$$P(\bar{x} < 6) = P(z < -3.65) = 0.000131$$

That is, if smoking during pregnancy actually does still lead to an average birth weight of 7 pounds, there was only a 0.000131 (or 0.0131%) chance of getting a sample mean as low as six or lower. This is an extremely unlikely event if the assumption is true. Therefore it is likely the assumption is not true.

## 2.3 Hypotheses Tests

## Statistical Hypotheses

- A hypothesis is a claim or statement about a property of a population.
  - Example: The population mean for systolic blood pressure is 120.
- A hypothesis test (or test of significance) is a standard procedure for testing a claim about a property of a population.
- Recall the example about birth weights with mothers who smoked during pregnancy.
  - Hypothesis: Smoking during pregnancy leads to an average birth weight of 7 pounds (the same as with mothers who do not smoke).

## Null and Alternative Hypotheses

• The null hypothesis is a statement that the value of a population parameter (such as the population mean) *is equal to* some claimed value.

 $-H_0: \mu = 7.$ 

- The **alternative hypothesis** is an alternative to the null hypothesis; a statement that says a parameter differs from the value given in the null hypothesis.
  - $-H_a: \mu < 7.$
  - $H_a: \mu > 7.$
  - $-H_a$ :  $\mu \neq 7$ .
- In hypothesis testing, assume the null hypothesis is true until there is strong statistical evidence to suggest the alternative hypothesis.
- Similar to an "innocent until proven guilty" policy.

#### Hypothesis tests

• (Many) hypothesis tests are all the same:

z or  $t=\frac{\text{sample statistic}-\text{null hypothesis value}}{\text{standard deviation of the sampling distribution}}$ 

- Example: hypothesis testing about  $\mu$ :
  - Sample statistic =  $\bar{x}$ .
  - Standard deviation of the sampling distribution of  $\bar{x}$ :

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

#### **P-values**

- Interpretation: *If the null hypothesis is correct,* than the p-value is the probability of obtaining a sample that yielded your statistic, or a statistic that provides even stronger evidence for the alternative hypothesis.
- The p-value is therefore a measure of *statistical significance*.
  - If p-values are small, there is sufficient statistical evidence in favor of the alternative hypothesis.
  - If p-values are large, there is insignificant statistical evidence. Therefore, you fail to reject the null hypothesis.
- Best practice is writing research: report the p-value, report your significance level (cut-off value), then reject / fail-to-reject.

# 3 Using R

## Using R

Online tutorial for first-time R user:http://tryr.codeschool.com/ Other resources:

- *R for Beginners*: PDF manual for learning R https://cran.r-project. org/doc/contrib/Paradis-rdebuts\_en.pdf
- An Introduction to R: PDF Reference Manual for common R tools https: //cran.r-project.org/doc/manuals/r-release/R-intro.pdf
- Google It! http://www.google.com
  - Usually useful results from Stackexchange.com or Stackoverflow.com