Stat 13, Intro. to Statistical Methods for the Life and Health Sciences.

- 1. Type I and type II errors.
- 2. Power.
- 3. Confidence Intervals for a proportion and the dog sniffing cancer example.

Start reading chapter 4.

http://www.stat.ucla.edu/~frederic/13/W23.

HW2 is due Fri Feb10 at 2pm to statgrader@stat.ucla.edu or statgrader2@stat.ucla.edu .

1. Type I and Type II errors

A type I error is a false positive.

Test says someone has a disease when they don't.

A type II error is a false negative.

Test says someone does not have a disease when they actually do.

Type I and Type II Errors

| TABLE 2.9 A summary of Type I and Type II errors | | | | | | | | | | |
|--|-------------------------------|-------------------------------|------------------------------------|--|--|--|--|--|--|--|
| | | What is true (unknown to us) | | | | | | | | |
| | | Null hypothesis is true | Null hypothesis is false | | | | | | | |
| What we decide | Reject null hypothesis | Type I error (false alarm) | Correct decision | | | | | | | |
| (based on data) | Do not reject null hypothesis | Correct decision | Type II error (missed opportunity) | | | | | | | |

The probability of a Type I error

- The significance level is the probability of a type I error, when Ho is true.
- Suppose the significance level is 0.05. If the null is true we would reject it 5% of the time and thus make a type I error 5% of the time.
- If you make the significance level lower, you have reduced the probability of making a type I error, but have increased the probability of making a type II error.

The probability of a Type II error

- The probability of a type II error is more difficult to calculate.
- In fact, the probability of a type II error is not even a fixed number. It depends on the value of the true parameter you are estimating.
- The probability of a type II error can be very high if:
 - The effect size is small.
 - The sample size is small.

2. Power

- The probability of rejecting the null hypothesis when it is false is called the **power** of a test.
- Power = 1 P(Type | I | error). It is usually expressed as a function of μ .
- We want a test with high power and this is aided by:
 - A large effect size, i.e. true μ far from the parameter in the null hypothesis.
 - A large sample size.
 - A small standard deviation.
 - A higher significance level means greater power.
 The downside is that you get more type I errors.

3. Estimation and confidence intervals.

Chapter 3

Chapter Overview

- So far, we can only say things like
 - "We have strong evidence that the long-run frequency of death within 30 days after a heart transplant at St. George's Hospital is greater than 15%."
 - "We do not have strong evidence kids have a preference between candy and a toy when trickor-treating."
- We want a method that says
 - "I believe 68 to 75% of all elections can be correctly predicted by the competent face method."

Confidence Intervals

- Interval estimates of a population parameter are called **confidence intervals**.
- We will find confidence intervals three ways.
 - Through a series of tests of significance to see which proportions are plausible values for the parameter.
 - Using the standard error (the standard deviation of the simulated null distribution) to help us determine the width of the interval.
 - Through traditional theory-based methods, i.e. formulas.

Statistical Inference: Confidence Intervals

Section 3.1

Section 3.1

Sonoda et al. (2011). Marine, a dog originally trained for water rescues, was tested to see if she could detect if a patient had colorectal cancer by smelling a sample of their breath.

- She first smells a bag from a patient with colorectal cancer.
- Then she smells 5 other samples; 4 from normal patients and 1 from a person with colorectal cancer
- She is trained to sit next to the bag that matches the scent of the initial bag (the "cancer scent") by being rewarded with a tennis ball.

In Sonoda et al. (2011). Marine was tested in 33 trials.

- Null hypothesis: Marine is randomly guessing which bag is the cancer specimen ($\pi = 0.20$)
- Alternative hypothesis: Marine can detect cancer better than guessing ($\pi > 0.20$)

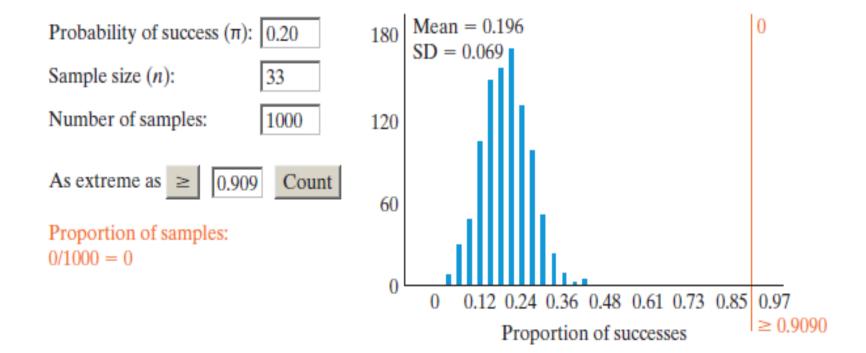
 π represents her long-run probability of identifying the cancer specimen.

- 30 out of 33 trials resulted in Marine correctly identifying the bag from the cancer patient
- So our sample proportion is

$$\hat{p} = \frac{30}{33} \approx 0.909$$

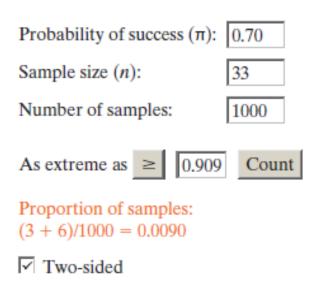
- Do you think Marine can detect cancer?
- What sort of p-value will we get?

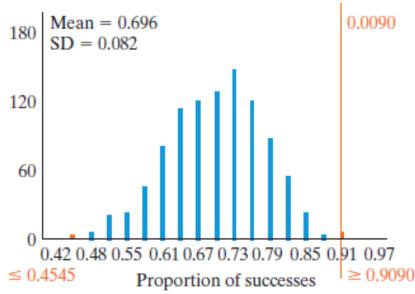
Our sample proportion lies more than 10 standard deviations above the mean and hence our p-value ~ 0.



- Can we estimate Marine's long run frequency of picking the correct specimen?
- Since our sample proportion is about 0.909, it is plausible that 0.909 is a value for this frequency. What about other values?
- Is it plausible that Marine's frequency is actually 0.70 and she had a lucky day?
- Is a sample proportion of 0.909 unlikely if $\pi = 0.70$?

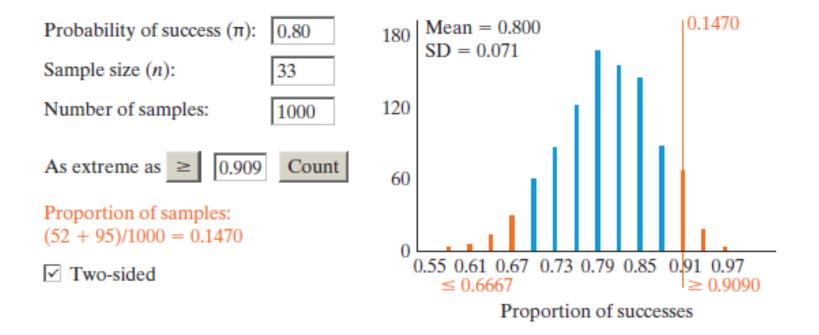
- H_0 : $\pi = 0.70$ H_a : $\pi \neq 0.70$
- We get a small p-value (0.0090) so we can essentially rule out 0.70 as her long run frequency.





- What about 0.80?
- Is 0.909 unlikely if π = 0.80?

- H_0 : $\pi = 0.80$ H_a : $\pi \neq 0.80$
- We get a large p-value (0.1470) so 0.80 is a
 plausible value for Marine's long-run frequency.



Developing a range of plausible values

- If we get a small p-value (like we did with 0.70) we will conclude that the value under the null is not plausible. This is when we reject the null hypothesis.
- If we get a large p-value (like we did with 0.80)
 we will conclude the value under the null is
 plausible. This is when we can't reject the
 null.

Developing a range of plausible values

- One could use software (like the one-proportion applet the book recommends) to find a range of plausible values for Marine's long term probability of choosing the correct specimen.
- We will keep the sample proportion the same and change the possible values of π .
- We will use 0.05 as our cutoff value for if a p-value is small or large. (Recall that this is called the significance level.)

 It turns out values between 0.761 and 0.974 are plausible values for Marine's probability of picking the correct specimen.

| Probability under null | 0.759 | 0.760 | 0.761 | 0.762 | | 0.973 | 0.974 | 0.975 | 0.976 |
|------------------------|-------|-------|-------|-------|---------|-------|-------|-------|-------|
| p-value | 0.042 | 0.043 | 0.063 | 0.063 | ••••• | 0.059 | 0.054 | 0.049 | 0.044 |
| Plausible? | No | No | Yes | Yes | Yes | Yes | Yes | No | No |

- (0.761, 0.974) is called a *confidence interval*.
- Since we used 5% as our significance level, this is a 95% confidence interval. (100% – 5%)
- 95% is the *confidence level* associated with the interval of plausible values.

- We would say we are 95% confident that Marine's probability of correctly picking the bag with breath from the cancer patient from among 5 bags is between 0.761 and 0.974.
- This is a more precise statement than our initial significance test which concluded Marine's probability was more than 0.20.
- Sidenote: We do not say $P\{\pi \text{ is in } (.761, .974)\} = 95\%$, because π is not random. The *interval* is random, and would change with a different sample. If we calculate an interval this way, then $P(\text{interval contains } \pi) = 95\%$.

Confidence Level

• If we increase the confidence level from 95% to 99%, what will happen to the width of the confidence interval?

- Since the confidence level gives an indication of how sure we are that we captured the actual value of the parameter in our interval, to be more sure our interval should be wider.
- How would we obtain a wider interval of plausible values to represent a 99% confidence level?
 - Use a 1% significance level in the tests.
 - Values that correspond to 2-sided p-values
 larger than 0.01 should now be in our interval.