Lecture 11

11.1 Calculating the Power of a Hypothesis Test

- It is relatively easy to calculate the power of the kinds of hypothesis tests we have been discussing.
- You'll recall that a test's statistical power is 1β . It is the probability that the test will falsely reject a positive result, or the probability of the commission of a Type II error.
- Go over handout called "Type I, Type II error."
- So how do we calculate β ?

$$eta = \Pr\left(\operatorname{Accept} H_0 | H_A \text{ true}\right)$$

$$= \Pr\left(\widehat{\theta} < \theta_0 + z_{\alpha} \sigma_{\widehat{\theta}} | \theta = \theta_A\right)$$

• Well we know that $\widehat{\theta}$ is distributed Normal with mean θ_A and standard deviation $\sigma_{\widehat{\theta}}$. So we therefore know the probability with which it will take on any value. First standardize $\widehat{\theta}$:

$$\begin{split} &= & \Pr\left(\frac{\widehat{\theta} - \theta_A}{\sigma_{\widehat{\theta}}} < \frac{\theta_0 + z_\alpha \sigma_{\widehat{\theta}} - \theta_A}{\sigma_{\widehat{\theta}}} \middle| \theta = \theta_A\right) \\ &= & \Phi\left(\frac{\theta_0 + z_\alpha \sigma_{\widehat{\theta}} - \theta_A}{\sigma_{\widehat{\theta}}}\right). \\ &= & \Phi\left(\frac{\theta_0 - \theta_A}{\sigma_{\widehat{\theta}}} + z_\alpha\right) \end{split}$$

• So a test's power is therefore

$$1-\Phi\left(rac{ heta_0- heta_A}{\sigma_{\widehat{ heta}}}+z_lpha
ight).$$

- Note that we have all the quantities necessary to calculate β and therefore power:
 - We've specified a θ_0 and θ_A .

- We've also specified a α and therefore a z_{α} .
- And we use our usual methods to obtain $\sigma_{\widehat{\theta}} = \frac{\sigma}{\sqrt{n}}$.
 - * (So we might actually write:

$$Power = 1 - \Phi\left(rac{ heta_0 - heta_A}{rac{\sigma}{\sqrt{n}}} + z_lpha
ight).$$

- Now for something interesting. What is the sign of $\frac{\partial \text{Power}}{\partial \alpha}$? Of $\frac{\partial \text{Power}}{\partial \sigma}$?
- NOTE TO SELF: Note that $\theta_0 \theta_A < 0$!
- Well, since Φ is monotonically increasing in its argument, we know that:
 - $-\frac{\partial z_{\alpha}}{\partial \alpha} < 0$, and so $\frac{\partial \Phi\left(\frac{\theta_0 \theta_A}{\sigma} + z_{\alpha}\right)}{\partial \alpha} < 0$, and so $\frac{\partial Power}{\partial \alpha} > 0$. As we increase α (that is, decrease our confidence coefficient) we increase power.
 - $-\frac{\partial \frac{\theta_0-\theta_A}{\sigma}}{\partial \sigma} = \frac{\partial \frac{(\theta_0-\theta_A)\sqrt{n}}{\sigma}}{\partial \sigma} > 0 \text{ (since } \theta_0-\theta_A<0), \text{ and so } \frac{\partial \Phi\left(\frac{\theta_0-\theta_A}{\sigma}+z_\alpha\right)}{\partial \sigma} > 0, \text{ and so } \frac{\partial \operatorname{Power}}{\partial \sigma}<0.$ As there is more variance in the population, our estimator becomes less precise, and the power of our statistical tests goes down.
 - Conversely, $\frac{\partial \frac{\theta_0 \theta_A}{\sigma}}{\partial n} < 0$, and so $\frac{\partial Power}{\partial n} > 0$.
 - And finally $\frac{\partial Power}{\partial(|\theta_0-\theta_A|)} = \frac{\partial Power}{\partial(\theta_A-\theta_0)}$ (since $\theta_0-\theta_A<0$). Since $\frac{\partial \Phi\left(\frac{\theta_0-\theta_A}{\sigma_N}+z_\alpha\right)}{\partial(\theta_A-\theta_0)}<0$, $\frac{\partial Power}{\partial(|\theta_0-\theta_A|)}>0$. The farther way you specify an alternative hypothesis away from the null, the less likely you are to falsely reject the null.
- A vice president in charge of sales for a large corporation claims that salespeople are averaging no more than 15 sales contacts per week. (He would like to increase this figure.) As a check on his claim, n = 36 salespeople are selected at random, and the number of contacts made by each is recorded for a single randomly selected week. The mean and variance of the 36 measurements were 17 and 9, respectively. Does the evidence contradict the vice president's claim? Use a test with level $\alpha = .05$.

$$- H_0 : \mu = 15$$

- $H_A : \mu > 15$
- We know that $\bar{Y} \sim \mathcal{N}(\mu_{\bar{Y}}, \sigma_{\bar{Y}})$, meaning our test statistic can be $Z = \frac{\bar{Y} \mu_0}{\sigma_{\bar{Y}}} = \frac{\bar{Y} \mu_0}{\frac{\sigma}{\sqrt{n}}}$
- Our rejection region for $\alpha = 0.05$ is given by z = 1.645.
- Thus $Z=\frac{\bar{Y}-\mu_0}{\frac{\sigma}{\sqrt{n}}}=\frac{17-15}{3/6}=4$ which is inside the rejection region.
- Now suppose the VP wants to be able to detect a difference equal to one call in the mean number of calls per week. That is, he wishes to test $H_0: \mu = 15$ against the alternative $H_A: \mu = 16$. What is the β for this test?
 - Given the same set-up as above, the rejection region for $\alpha = .05$ is $\bar{y} > \mu_0 + 1.645 \left(\frac{\sigma}{\sqrt{n}}\right)$.
 - Substituting our values, we get $\bar{y} > 15 + 1.645 \left(\frac{3}{\sqrt{36}}\right)$ or $\bar{y} > 15.8225$.
 - Thus $\beta = P(\bar{Y} \le 15.8225 | \mu_A = 16)$ or $P\left(\frac{\bar{Y} \mu_a}{\sigma/\sqrt{n}} \le \frac{15.8225 16}{3/\sqrt{36}}\right) = P(Z \le -.36) = .3594$

11.2 Another way to report results of a statistical test: *p*-values

- You'll recall that *α*, the probability of a Type I ("false positive") error associated with a statistical test, is often called the test's **significance level**.
- In the hypothesis testing regime we've discussed so far, we:
 - pick a significance level
 - determine the critical value(s) of the test statistic associated with the significance level
 - and then determine whether to accept or reject the null hypothesis by comparing our test statistic with the critical value.
- This is all very black-or-white: the result is either significant or it's not. But if report only
 whether we reject or accept the null, we're actually failure to report a fair amount of information.
- Another way to report results of a statistical test that provides the reader with this additional
 information is to report what's called its *p*-value.

- For any test statistic, the *p-value*, or **attained significance level**, is the smallest level of significance α for which the observed data indicate that the null hypothesis should be rejected.
- The smaller a *p*-value is, the more compelling is the evidence that the null hypothesis should be rejected. That is because the null should be rejected for any value of *α down to and including* the *p*-value.
- Reporting the *p*-value permits the reader to make her own choice about whether the observed data justify a rejection of the null.
- This should be obvious, but:

Reject
$$H_0 \iff p \le \alpha$$

Accept
$$H_0 \iff \alpha \leq p$$

– A psychological study was conducted to compare the reaction times of men and women to a stimulus. Independent random samples of 50 men and 50 women were employed in the experiment. The results are shown in the table below. Do the data present sufficient evidence to suggest a difference between true mean reaction times for men and women? Use $\alpha = .05$.

Men Women

$$n_1 = 50$$
 $n_2 = 50$
 $\bar{y}_1 = 3.6$ $\bar{y}_2 = 3.8$
 $s_1^2 = .18$ $s_2^2 = .14$

$$- H_0 : (\mu_1 - \mu_2) = 0$$

-
$$H_A: (\mu_1 - \mu_2) \neq 0$$

– Test statistic:
$$Z=rac{(ar{Y}_1-ar{Y}_2)- heta_0}{\sqrt{rac{\sigma_1^2}{n_1}+rac{\sigma_2^2}{n_2}}}$$

- Setting $\theta_0 = 0$ yields $Z = \frac{(3.6-3.8)-0}{\sqrt{\frac{.18}{50} + \frac{.14}{50}}} = -2.5$ which is below the $-z_{\alpha/2} = -1.96$ so we reject the null.
- What is the p-value associated with this conclusion?

– $P(Z \le -2.5) = .0062$ which means $\alpha = 0.0062*2 = 0.0124$. Note that, if we set $\alpha = .01$, we would NOT be able to reject.