Unit 5: Inference for categorical data

2. Comparing two proportions

Sta 101 - Spring 2015

Duke University, Department of Statistical Science

March 18, 2015

2. Main ideas

- 1. CLT also describes the distribution of $\hat{p}_1 \hat{p}_2$
- 2. For HT where $H_0: p_1 = p_2$, pool!
- 3. When S-F fails, simulate!

Applications

- 1. Two population proportions, small sample
- Single population proportion, small sample

- MT2 Review Monday, March 23, 7-8pm
- OH next week Monday and Tuesday 3-5pm
- MT review materials to be posted on Sakai over the weekend
- RA5 opens on Sunday (but will also cover material from Monday's class so you might want to wait to take it) and will close at midnight on Monday
- ► Mao's office hours moved to tomorrow 7-9pm, Tori is there tonight 7-9pm.

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CLT also describes the distribution of $\hat{p}_1 - \hat{p}_2$

$$(\hat{p}_1 - \hat{p}_2) \sim N \left(mean = (p_1 - p_2), SE = \sqrt{\frac{p_1(1 - p_1)}{n_1} + \frac{p_2(1 - p_2)}{n_2}} \right)$$

Conditions:

- ▶ Independence: Random sample/assignment + 10% rule
- ▶ Sample size / skew: At least 10 successes and failures

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As with working with a single proportion,

- ▶ When doing a HT where $H_0: p_1 = p_2$ (almost always for HT), use expected counts / proportions for S-F condition and calculation of the standard error.
- Otherwise use observed counts / proportions for S-F condition and calculation of the standard error.

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Expected proportion of success for both groups when $H_0: p_1 = p_2$ is defined as the *pooled proportion*:

$$\hat{p}_{pool} = \frac{total\ successes}{total\ sample\ size} = \frac{suc_1 + suc_2}{n_1 + n_2}$$

Suppose in group 1 30 out of 50 observations are successes, and in group 2 20 out of 60 observations are successes. What is the pooled proportion?

- (a) $\frac{30}{50}$
- (b) $\frac{20}{60}$
- (c) $\frac{30}{50} + \frac{20}{60}$
- (d) $\frac{30+20}{50+60}$
- (e) $\frac{\frac{30}{50} + \frac{20}{60}}{2}$

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When S-F fails, simulate!

- ▶ If the S-F condition is met, can do theoretical inference: Z test, Z interval
- ▶ If the S-F condition is not met, must use simulation based methods: randomization test, bootstrap interval

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"Healthy adults immunized with an experimental malaria vaccine, called PfSPZ may be completely protected from infection, according to government researchers." reported Time magazine in Aug 2013. The vaccine contains weakened forms of the live parasite -- *Plasmodium falciparum* -- responsible for causing malaria. In a randomized trial, none of the six patients who received the vaccine developed malaria, while five of the six who were not vaccinated became infected. Do these data provide convincing evidence of a difference in rate of malaria infection?

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		Malaria	No malaria	
Group	Vaccine	0	6	6
	No vaccine	5	1	6
	Total	5	7	12

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 No malaria

 Vaccine
 0
 6
 6

 No vaccine
 5
 1
 6

 Total
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 7
 12

$$H_0: p_T = p_C \qquad H_A: p_T \neq p_C$$

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$$H_0: p_T = p_C \qquad H_A: p_T \neq p_C$$

Conditions:

- 1. Independence: Patients are randomly assigned to treatment groups
- 2. Success-failure: ?

Assuming that the null hypothesis $(H_0: p_T = p_C)$ is true, which of the following is the pooled proportion of patients with malaria in the two groups?

(a)
$$\frac{6}{12} = 0.5$$

(b)
$$\frac{5}{12} = 0.417$$

(c)
$$\frac{0}{5} = 0$$

(d)
$$\frac{6}{7} = 0.857$$

(e)
$$\frac{7}{12} = 0.583$$

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Assuming that the null hypothesis $(H_0: p_T = p_C)$ is true, how many patients would we expect to get infected with malaria in the vaccine group?

(a)
$$0.417 \times 12 = 5$$

(b)
$$0.417 \times 6 = 2.5$$

(c)
$$0.417 \times 5 = 2.085$$

(d)
$$0.5 \times 6 = 3$$

(e)
$$0.583 \times 12 = 7$$

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$$\hat{p}_{pool} = 5/12 = 0.417$$
$$1 - 0.417 = 0.583$$

$$Exp \ S_T = 0.417 \times 6 = 2.5$$
 $Exp \ S_C = 0.417 \times 6 = 2.5$ $Exp \ F_T = 0.583 \times 6 = 3.5$ $Exp \ F_C = 0.583 \times 6 = 3.5$

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- Calculate the difference between the proportions of "malaria" in the vaccine and no vaccine decks, and record this number.

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- Calculate the difference between the proportions of "malaria" in the vaccine and no vaccine decks, and record this number.
- Repeat steps (3) and (4) many times to build a randomization distribution of differences in simulated proportions.

Simulate in R

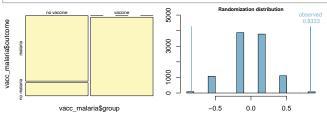
```
download("https://stat.duke.edu/-mc301/data/vacc_malaria.csv", destfile = "vacc_malaria.csv")
vacc_malaria = read.csv("vacc_malaria.csv")
inference(vacc_malaria$outcome, vacc_malaria$group, success = "malaria", est = "proportion",
    type = "ht", null = 0, alternative = "twosided", method = "simulation", seed = 1028)
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```
Response variable: categorical, Explanatory variable: categorical
Difference between two proportions -- success: malaria
Summary statistics:

x
y
no vaccine vaccine Sum
malaria
5
0
5
no malaria
1
6
7
Sum
6
6
12
Observed difference between proportions (no vaccine-vaccine) = 0.8333
HO: p_no vaccine - p_vaccine = 0
HA: p_no vaccine - p_vaccine != 0
p-value = 0.0152
```



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Application exercise: App Ex 5.2

See course website for details.

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Summary of main ideas

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