

Math 362: Mathematical Statistics II

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Chapter 9. Two-Sample Inferences

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Plan

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§ 9.5 Confidence Intervals for the Two-Sample Problem

By the central limit theorem, when n and m are large

$$\frac{\frac{X}{n} - \frac{Y}{m} - \mathbb{E}\left(\frac{X}{n} - \frac{Y}{m}\right)}{\sqrt{\text{Var}\left(\frac{X}{n} - \frac{Y}{m}\right)}} \underset{\text{approx.}}{\sim} \mathcal{N}(0, 1)$$

Under $H_0 : p_X = p_Y$,

$$\mathbb{E}\left(\frac{X}{n} - \frac{Y}{m}\right) = 0$$

$$\text{Var}\left(\frac{X}{n} - \frac{Y}{m}\right) = \frac{p(1-p)}{n} + \frac{p(1-p)}{m}$$

The MLE for p under H_0 is

$$\hat{p}_e = \frac{x+y}{n+m}$$

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Testing $H_0 : p_X = p_Y$

v.s.

(at the α level of significance)

$$z = \frac{\frac{x}{n} - \frac{y}{m}}{\sqrt{p_e(1 - p_e) \left(\frac{1}{n} + \frac{1}{m}\right)}}, \quad p_e = \frac{x + y}{n + m}$$

$H_1 : p_X < p_Y:$

Reject H_0 if

$$z \leq -z_\alpha$$

$H_1 : p_X \neq p_Y:$

Reject H_0 if

$$|z| \geq z_{\alpha/2}$$

$H_1 : p_X > p_Y:$

Reject H_0 if

$$z \geq z_\alpha$$

E.g. Nightmares among men and women:

Is 34.4% significantly different from 31.1% ($\alpha = 0.05$)?

Sol. ...



E.g. Nightmares among men and women:

	Men	Women	Total
Nightmares often	55	60	115
Nightmares seldom	105	132	237
Totals	160	192	
% often:	34.4	31.3	

Is 34.4% significantly different from 31.1% ($\alpha = 0.05$)?

Sol. ...

