

# STAT 135 Lecture 11

Henry Liev

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## Lemma 0.1 (Neyman Pearson Lemma)

Fill cart with items that are the best deal  $\frac{P(X|H_1)}{P(X|H_0)}$ ,  $P(X|H_1)$  is the value and  $P(X|H_0)$  is the price, until you run out of budget. Maximizes the power  $(1 - \beta)$  for your  $\alpha$ .

## Example 0.2 (Duality of Confidence Intervals and Hypothesis Tests)

Let  $X_1, \dots, X_n \stackrel{iid}{\sim} \mathcal{N}(\mu, 1)$

Testing:  $H_0 : \mu = \mu_0$

$H_1 : \mu \neq \mu_0$

**Solution.** Get a level 0.05 test by reject when  $|\bar{X} - \mu_0| > \frac{1.96}{\sqrt{n}}$   
Accept when  $\frac{-1.96}{\sqrt{n}} \leq \mu_0 - \bar{X} \leq \frac{1.96}{\sqrt{n}} \rightarrow \bar{X} - \frac{1.96}{\sqrt{n}} \leq \mu_0 \leq \bar{X} + \frac{1.96}{\sqrt{n}}$

$\{\mu : \mu \in 95\% CI\} = \{\mu : H_0 : \mu = \mu_0 \text{ not rejected}\}$

Set of  $\mu$  in CI is the same as set of  $\mu$  which would not be rejected testing From 95% CI  $\Rightarrow$  Hypothesis test, reject outside the interval

From Hypothesis test  $\Rightarrow$  CI (Those null hypothesis wouldn't be rejected)  $\Rightarrow$  Can work back and forth between CI and Hypothesis Testing ■

## Remark 0.3 (Comparing two samples)

$X_1, \dots, X_n \stackrel{iid}{\sim} f, \quad Y_1, \dots, Y_m \stackrel{iid}{\sim} g$

Applications:

- Treatment vs control
- A/B Testing

## Example 0.4

$X_1, \dots, X_n \stackrel{iid}{\sim} \mathcal{N}(\mu_X, \sigma^2)$

$Y_1, \dots, Y_m \stackrel{iid}{\sim} \mathcal{N}(\mu_Y, \sigma^2)$

$\bar{X} - \bar{Y} \sim \mathcal{N}(\mu_Y - \mu_X, \sigma^2(\frac{1}{n} + \frac{1}{m}))$

$\frac{\bar{X} - \bar{Y} - (\mu_X - \mu_Y)}{\sigma \sqrt{\frac{1}{n} + \frac{1}{m}}} \sim N(0, 1)$

$\bar{X} - \bar{Y} \pm 1.96 SE(\bar{X} - \bar{Y}) = \bar{X} - \bar{Y} \pm 1.96 \sigma \sqrt{\frac{1}{n} + \frac{1}{m}}$

Reject  $\mu_X = \mu_Y$  when  $\mu_X - \mu_Y \neq 0$

### Method 0.5 (One sample t-test)

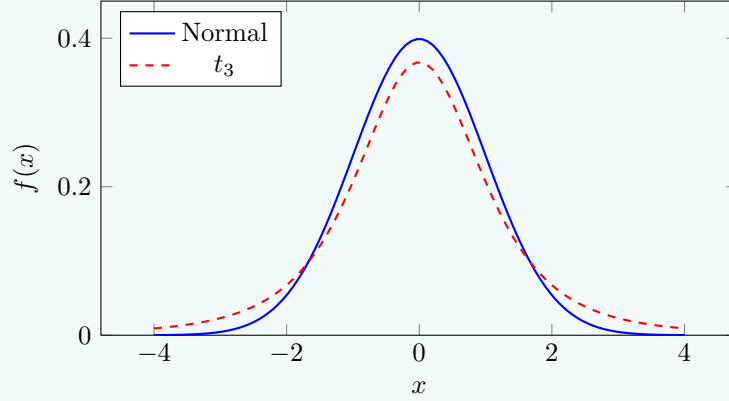
$X_1, \dots, X_n \stackrel{iid}{\sim} \mathcal{N}(\mu, \sigma^2)$  with  $\sigma^2$  known

$\bar{X} \pm 1.96 \frac{\sigma}{\sqrt{n}}$  Exact 95%

$\sigma$  unknown

If we estimate  $\sigma^2$  by  $s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$

Then  $\frac{\sqrt{n}(\bar{X} - \mu)}{s} \sim t_{n-1}$



If we want a  $(1 - \alpha)\%$  Confidence interval for  $\mu$ .

$$\bar{X} \pm t_{n-1}(\frac{\alpha}{2}) \frac{s}{\sqrt{n}}, \quad t_{n-1}(\frac{\alpha}{2}) > 1.96$$

### Method 0.6 (Back to two sample problem)

$X_1, \dots, X_n \stackrel{iid}{\sim} \mathcal{N}(\mu_X, \sigma^2)$

$Y_1, \dots, Y_m \stackrel{iid}{\sim} \mathcal{N}(\mu_Y, \sigma^2)$

$\sigma^2$  unknown

$$\frac{(\bar{X} - \bar{Y}) - (\mu_X - \mu_Y)}{S_p \sqrt{\frac{1}{n} + \frac{1}{m}}} \sim t_{n+m-2}$$

$$S_p^2 = \frac{(n-1)S_X^2 + (m-1)S_Y^2}{n+m-2} \quad \text{“Pooled Estimate”}$$

$$S_X^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$$S_Y^2 = \frac{1}{m-1} \sum_{i=1}^m (y_i - \bar{y})^2$$

$$E(S_p^2) = \frac{(n-1)E(S_X^2) + (m-1)E(S_Y^2)}{n+m-2} = \frac{(n-1+m-1)\sigma^2}{n+m-2} = \sigma^2$$

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**Method 0.7** (Inference for two sample problem)

Exact  $(1 - \alpha)$  CI for  $\mu_X - \mu_Y$

$$\bar{X} - \bar{Y} \pm t_{n+m-2}(\frac{\alpha}{2}) S_p \sqrt{\frac{1}{n} + \frac{1}{m}}$$

$$H_0 : \mu_X = \mu_Y$$

$$H_1 : \mu_X \neq \mu_Y$$

Level  $\alpha$  test: Reject for large  $\frac{|\bar{X} - \bar{Y}|}{S_p \sqrt{\frac{1}{n} + \frac{1}{m}}}$

**Definition 0.8**

$$H_0 : \mu_X = \mu_Y$$

One-Sided Hypothesis:

$$H_1 : \mu_X > \mu_Y$$

$$H_1 : \mu_X < \mu_Y$$

Two-Sided Hypothesis

$$H_1 : \mu_X \neq \mu_Y$$

Use two-sided test unless you know direction a priori

Takeaways:

- Newyman Pearson Lemma Intuition
- Duality of CI/Hypothesis Tests
- Two Sample Problem
- Solved it for normally distributed data with common unknown variance