

- ▶ Unit 1: Introduction to Probability
- ▶ Unit 2: Probability Distributions
- ▶ **Unit 3: Statistical Inference**
- ▶ Unit 4: Introduction to Linear Regression
- ▶ Unit 5: Regression Analysis
- ▶ Unit 6: Regression Modeling

Announcements

► Past:

Unit 2 Individual Assignment solutions available and scores posted.

► Present:

Unit 3 Team Assignment **due Monday, 23:59 ET** (Durham local time).

► Future:

Unit 4 materials and individual assignment available.

► **Confidence Intervals** $PE \pm cv \times SE$

PE : Point Estimate (e.g., observed sample mean)

SE : Standard Error (e.g., observed sample std.deviation divided by the sqrt of sample size: $SE = S/\sqrt{n}$)

cv : Critical Value is determined by a desired confidence level

- 68% confidence interval: $PE \pm 1SE$
- 95% confidence interval: $PE \pm 2SE$
- 99.7% confidence interval: $PE \pm 3SE$

Can compute precisely for any other confidence level

Wider confidence interval (less informative) \iff Higher confidence level (more reliable)

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► **Estimating the (unknown) population mean**

$PE =$ observed sample mean \bar{x} , $SE = S/\sqrt{n}$

► **Estimating the (unknown) population proportion**

$PE =$ observed sample proportion \hat{p} , $SE = \sqrt{\hat{p}(1 - \hat{p})/n}$

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► Hypothesis Testing (supplementary)

$1 - (\text{p-value})$: largest confidence level in a claim, based on observed sample data

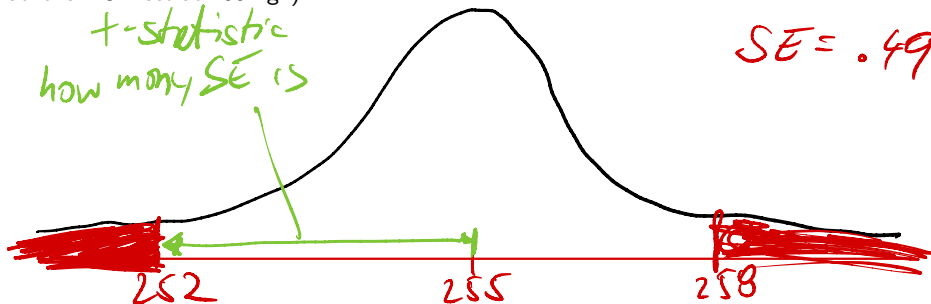
p-value: probability that the observed sample data is “extreme” and led to making a false claim

Production process calibration (revisited)

A manufacturer of a generic drug produces caplets which contain 250mg of the medication substance. The production process includes a natural slight variability in the exact contents of each caplet, so the company set a mean weight target at 255mg per caplet. (This allowable 2% increase in the target mean ensures sufficient medication in each caplet, without risking overdosage.)

A sample of 150 caplets is selected from the output of the caplet production process. Each caplet is carefully inspected for its medication content, which is recorded in the data file CapletWeights_Data.

Based on the observed sample data how confident can one be that the mean weight target is not set correctly (i.e., that it is NOT set at 255mg?)



The p-value

- ▶ The p-value quantifies uncertainty about the correctness of the claim/decision/conclusion based on the observed sample data.

The p-value can be interpreted as the probability of making a mistake due to observed sample being so extreme that the statistical analysis of sample data yield to the incorrect claim/decision/conclusion.

- ▶ **Want p-value small**

How small should p-value be? It depends on the business problem.

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- ▶ Hypothesis Testing

A way to frame sample-based statistical analysis when determining correctness of a specific claim.

Conclusions from any sample-based statistical analysis have inherent uncertainty (due to sampling)

The p-value is the probability that hypothesis testing lead to an incorrect claim.

$1 - (\text{p-value})$ is the confidence level that the claim is correct.

Efficacy of Moderna mRNA-1273 SARS-CoV-2 Vaccine

	mRNA-1273 Vaccine N=14,550	Placebo N=14,598
Symptomatic Covid-19	11	185
Severe Covid-19	0	30

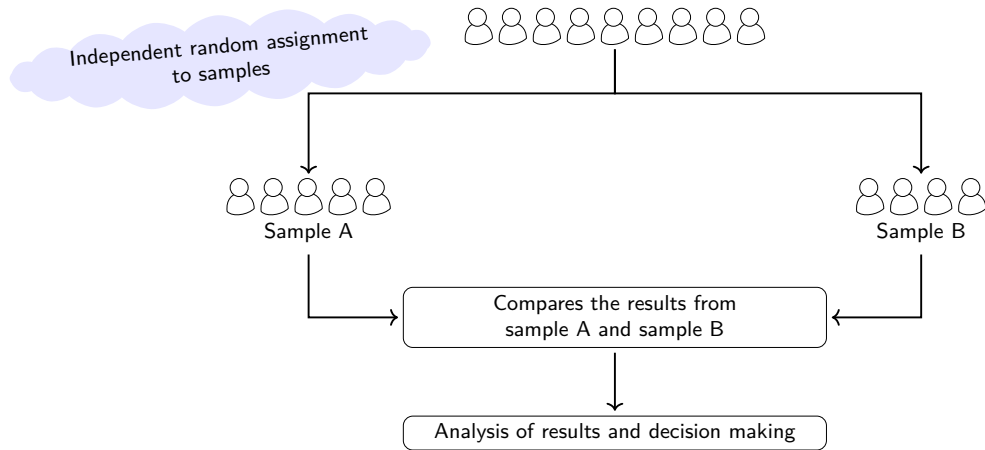
Vaccine efficacy of 94.1% (95% CI, 89.3–96.8%; $P < 0.001$)

“the trial was designed for the null hypothesis that the efficacy of the mRNA-1273 vaccine is 30% or less”

(FDA requirement: *“point estimate..at least 50%”* and *“lower bound of..confidence interval..is $>30\%$ ”*)

- Want to claim that the vaccine has efficacy larger than 30% in the population (i.e., vaccination reduces the incidence of symptomatic disease by more than 30%)
 - Want to reject the null hypothesis
- Having observed 94.1% efficacy in the trial, what is the probability that the efficacy is 30% or less?

A/B testing: comparing independent samples



Note. The two samples might also be referred to as the *control* and *treatment* (or experimental) samples (or groups)

A/B testing formally

Question: Is $p_A \neq p_B$? In other words, is difference $d = p_A - p_B$ different from zero?

Sample A

- ▶ Sample A size: n_A
- ▶ Sample A proportion: \hat{p}_A
- ▶ Standard error for \hat{p}_A : $\sqrt{\frac{\hat{p}_A(1-\hat{p}_A)}{n_A}}$

Sample B

- ▶ Sample B size: n_B
- ▶ Sample B proportion: \hat{p}_B
- ▶ Standard error for \hat{p}_B : $\sqrt{\frac{\hat{p}_B(1-\hat{p}_B)}{n_B}}$

A/B testing formally

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Sample B

- ▶ Sample B size: n_B
- ▶ Sample B proportion: \hat{p}_B
- ▶ Standard error for \hat{p}_B : $\sqrt{\frac{\hat{p}_B(1-\hat{p}_B)}{n_B}}$

- ▶ Compute the point estimate for the difference of two sample proportions $\hat{d} = \hat{p}_A - \hat{p}_B$
- ▶ Compute the standard error for the difference of two sample proportions $SE = \sqrt{\frac{\hat{p}_A(1-\hat{p}_A)}{n_A} + \frac{\hat{p}_B(1-\hat{p}_B)}{n_B}}$
- ▶ Compute t-statistic $= \hat{d}/SE$, and determine the p-value using the two-sided t-test

Unit 3 Team Assignment

- ▶ Due Monday, 23:59 ET (Durham local time).
- ▶ Five questions with different point allocations per question
- ▶ Need to provide support for your answers. Communicating effectively is important.
- ▶ One submission per team: two-page pdf document (no other documents)

Assignment questions:

- ▶ Questions 1-2: confidence intervals
- ▶ Question 3-5: analytics supported business decisions