

Lecture 16: Hypothesis Testing (III)

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Calculus-based Introduction to Statistics (S1201)

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COLUMBIA UNIVERSITY
IN THE CITY OF NEW YORK

Two-sided one-sample z-test

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} N(\mu, \sigma^2)$ with σ^2 **known**. Our hypothesis testing problem is

$$H_0 : \mu = \mu_0 \quad \text{v.s.} \quad H_a : \mu \neq \mu_0,$$

for some constant μ_0 . We want to control Type-I error rate $\leq \alpha$.

- Test statistic $Z = \frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma} \sim N(0, 1)$
- The test: **reject H_0 if $|Z| > z_{\alpha/2}$, otherwise accept H_0** . This is called the **two-sided one-sample z-test** with level α .
- **p-value** = $\mathbb{P}(|Z| > |z|) = 2\Phi\left(-\left|\frac{\sqrt{n}(\bar{x} - \mu_0)}{\sigma}\right|\right)$
- **When $\mu = \mu_1$, $\bar{X} \sim N(\mu_1, \sigma^2/n)$** . Type-II error rate
$$\begin{aligned} &= \mathbb{P}(-z_{\alpha/2} \leq Z = \frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma} \leq z_{\alpha/2}) \\ &= \mathbb{P}\left(\frac{\sqrt{n}(\mu_0 - \mu_1)}{\sigma} - z_{\alpha/2} \leq \frac{\sqrt{n}(\bar{X} - \mu_1)}{\sigma} \leq \frac{\sqrt{n}(\mu_0 - \mu_1)}{\sigma} + z_{\alpha/2}\right) \\ &= \Phi\left(\frac{\sqrt{n}(\mu_0 - \mu_1)}{\sigma} + z_{\alpha/2}\right) - \Phi\left(\frac{\sqrt{n}(\mu_0 - \mu_1)}{\sigma} - z_{\alpha/2}\right) \end{aligned}$$

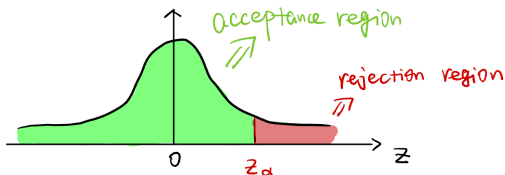
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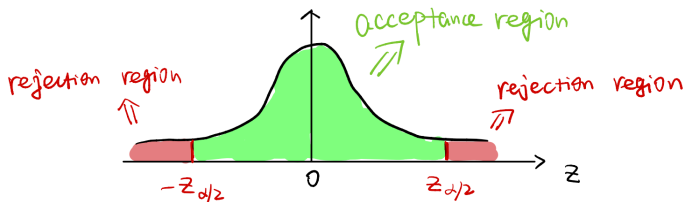
for some constant μ_0 . We want to control Type-I error rate $\leq \alpha$.

- Test statistic $Z = \frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma} \sim N(0, 1)$
- The test: **reject H_0 if $Z > z_\alpha$, otherwise accept H_0** . This is called the **one-sided one-sample z-test** with level α .
- **p-value** = $\mathbb{P}(Z > z) = 1 - \Phi\left(\frac{\sqrt{n}(\bar{x} - \mu_0)}{\sigma}\right)$
- **When $\mu = \mu_1$, $\bar{X} \sim N(\mu_1, \sigma^2/n)$. Type-II error rate**
 $= \mathbb{P}(Z = \frac{\sqrt{n}(\bar{X} - \mu_1)}{\sigma} \leq z_\alpha) = \mathbb{P}\left(\frac{\sqrt{n}(\bar{X} - \mu_1)}{\sigma} \leq \frac{\sqrt{n}(\mu_0 - \mu_1)}{\sigma} + z_\alpha\right)$
 $= \Phi\left(\frac{\sqrt{n}(\mu_0 - \mu_1)}{\sigma} + z_\alpha\right)$

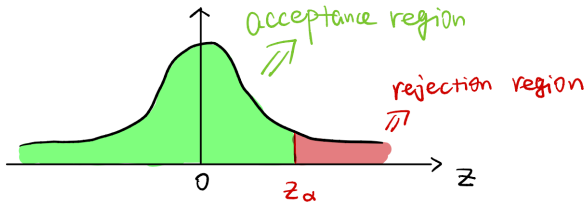


Compare two-sided and one-sided z-tests

Two-sided z-test: H_a includes both directions ($\mu \neq \mu_0$)

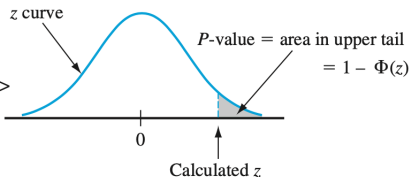


One-sided z-test: H_a includes only one direction ($\mu > \mu_0$ or $\mu < \mu_0$)

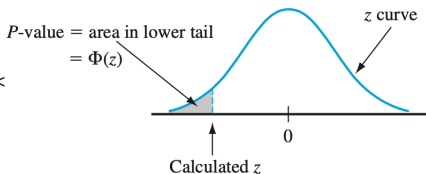


p-value in two-sided and one-sided z-tests

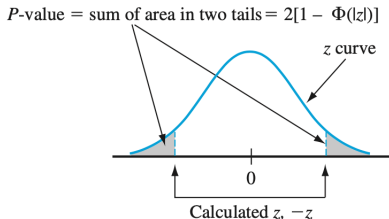
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 H_a contains the inequality $>$



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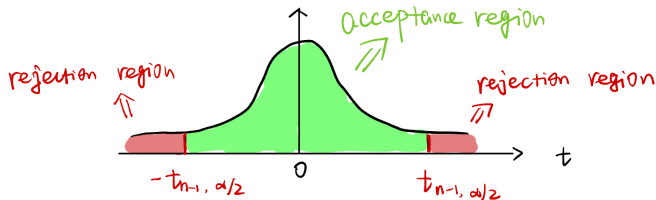
Two-sided one-sample t-test

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} N(\mu, \sigma^2)$ with σ^2 **unknown**.
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$$H_0 : \mu = \mu_0 \quad \text{v.s.} \quad H_a : \mu \neq \mu_0,$$

for some constant μ_0 . We want to control Type-I error rate $\leq \alpha$.

- Test statistic $T = \frac{\sqrt{n}(\bar{X} - \mu_0)}{S} \sim t_{n-1}$
- The test: **reject H_0 if $T < -t_{n-1, \alpha/2}$ or $T > t_{n-1, \alpha/2}$, otherwise accept H_0** . This is called the **two-sided one-sample z-test** with level α .
- **p-value** = $\mathbb{P}(|T| > |t|) = 2F\left(-\left|\frac{\sqrt{n}(\bar{x} - \mu_0)}{s}\right|\right)$, F : cdf of t_{n-1} -distribution
- Type-II error rate is a bit harder to calculate in this case so we won't discuss it



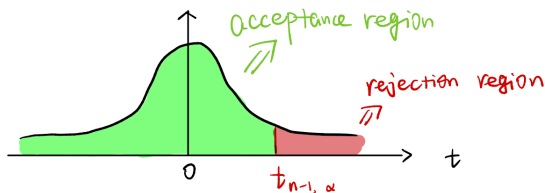
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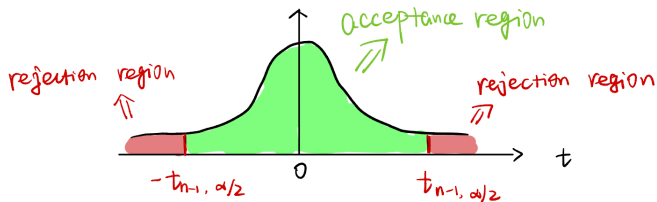
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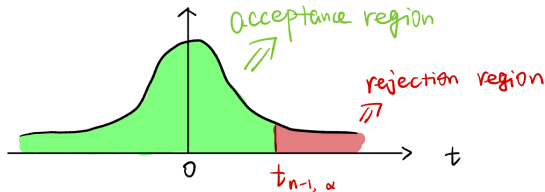


Compare two-sided and one-sided t-tests

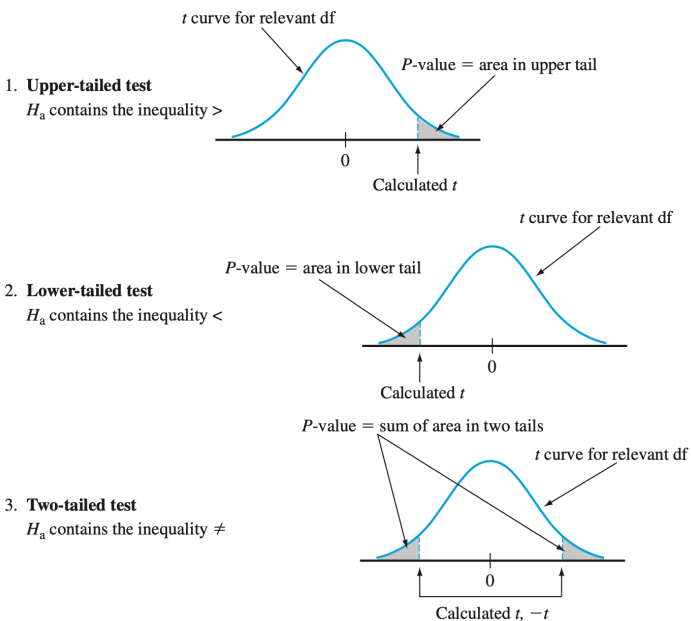
Two-sided t-test: H_a includes both directions ($\mu \neq \mu_0$)



One-sided t-test: H_a includes only one direction ($\mu > \mu_0$ or $\mu < \mu_0$)



p-value in two-sided and one-sided t-tests



Summary of testing mean of normal distribution

- σ known, $H_0 : \mu = \mu_0$
 - ▷ $H_a : \mu \neq \mu_0 \Rightarrow$ two-sided z-test
 - ▷ $H_a : \mu > \mu_0$ (or $\mu < \mu_0$) \Rightarrow one-sided z-test
- σ unknown, $H_0 : \mu = \mu_0$
 - ▷ $H_a : \mu \neq \mu_0 \Rightarrow$ two-sided t-test
 - ▷ $H_a : \mu > \mu_0$ (or $\mu < \mu_0$) \Rightarrow one-sided t-test

Today's goal

- Know how to use z-test to test the mean of general distributions (valid only when n is very large, hence called "large-sample z-test")
- Understand two-sided and one-sided hypothesis testing procedure for the population proportion p (i.e. the parameter p in Bernoulli(p) or Bin(n, p), valid only when n is very large, also called "large-sample z-test")
- Know how to calculate Type-I, Type-II and p-value under the aforementioned scenarios

Large-sample z-test for Mean of General Distributions

Example (the last time): fire protection system

A manufacturer of sprinkler systems used for fire protection in office buildings claims that the true average system-activation temperature is 130 °F. A sample of $n = 90$ systems, when tested, yields a sample average activation temperature of 131.08 °F. If the sample standard deviation of activation times is 2.25 °F **[with NO normal distribution!]**, does the data contradict the manufacturer's claim at significance level $\alpha = 0.05$?



H_0 : average activation temperature $\mu = 130$ °F v.s. H_a : $\mu \neq 130$ °F

Recall the general procedure to construct a test

- (1) Construct the null and alternative hypotheses. ✓
- (2) Given the level α , develop a test that satisfies Type-1 error rate $\mathbb{P}(\text{reject } H_0 | H_0 \text{ true}) \leq \alpha$. ?

How to do step (2):

- Construct a test statistic $T(X_1, \dots, X_n)$, which is a function of samples X_1, \dots, X_n and whose distribution under H_0 is **known** ✗ **We don't know the distribution! --- But we may use CLT!**
- Find a random region B such that Type-1 error rate $\mathbb{P}(\text{reject } H_0 | H_0 \text{ true}) = \mathbb{P}(T(X_1, \dots, X_n) \in B) = \alpha$
- Then the rejection region is B , and the corresponding test is
$$\begin{cases} \text{reject } H_0, & \text{if } T(X_1, \dots, X_n) \in B, \\ \text{accept } H_0, & \text{if } T(X_1, \dots, X_n) \notin B. \end{cases}$$

Two-sided large-scale z-test: derivation

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim}$ **some unknown distribution** with both mean μ and standard deviation σ^2 **unknown**.

Our hypothesis testing problem is

$$H_0 : \mu = \mu_0 \quad \text{v.s.} \quad H_a : \mu \neq \mu_0,$$

for some constant μ_0 . We want to control Type-I error rate $\leq \alpha$.

Step 1: Under H_0 , **by CLT and Law of Large Numbers**,

$$\bar{X} \stackrel{d}{\approx} N(\mu_0, \sigma^2/n), \quad S \approx \sigma^2 \Rightarrow Z = \frac{\sqrt{n}(\bar{X} - \mu_0)}{S} \approx \frac{\sqrt{n}(\bar{X} - \mu_0)}{\sigma} \stackrel{d}{\approx} N(0, 1)$$

Step 2: The **larger** $|Z|$ is, more likely H_a is true. Therefore the rejection region would be in the form $|Z| > c$. Note that

$$\mathbb{P}(|Z| > z_{\alpha/2}) \approx \alpha.$$

Step 3: The rejection region is $|Z| > z_{\alpha/2}$, i.e. $\bar{X} < \mu_0 - \frac{S}{\sqrt{n}} z_{\alpha/2}$ or

$$\begin{aligned} \bar{X} > \mu_0 + \frac{S}{\sqrt{n}} z_{\alpha/2}. \quad \text{p-value} &= \mathbb{P}(|Z| > |z|) = \mathbb{P}\left(|Z| > \left| \frac{\sqrt{n}(\bar{x} - \mu_0)}{s} \right| \right) \\ &\approx 2\Phi\left(- \left| \frac{\sqrt{n}(\bar{x} - \mu_0)}{s} \right| \right) \end{aligned}$$

Two-sided large-scale z-test

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim}$ **some unknown distribution** with both mean μ and standard deviation σ^2 **unknown**.

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- The test: **reject H_0 if $|Z| > z_{\alpha/2}$, otherwise accept H_0** . This is called the **two-sided large-sample (one-sample) z-test** with level α .
- **p-value** = $\mathbb{P}(|Z| > |z|) \approx 2\Phi\left(-\left|\frac{\sqrt{n}(\bar{x} - \mu_0)}{s}\right|\right)$
- **When $\mu = \mu_1$, $\bar{X} \stackrel{d}{\approx} N(\mu_1, \sigma^2/n) \Rightarrow Z' = \frac{\sqrt{n}(\bar{X} - \mu_1)}{S} \stackrel{d}{\approx} N(0, 1)$** . Type-II error rate = $\mathbb{P}(-z_{\alpha/2} \leq Z = \frac{\sqrt{n}(\bar{X} - \mu_0)}{S} \leq z_{\alpha/2})$
 $\approx \mathbb{P}\left(\frac{\sqrt{n}(\mu_0 - \mu_1)}{S} - z_{\alpha/2} \leq \frac{\sqrt{n}(\bar{X} - \mu_1)}{S} \leq \frac{\sqrt{n}(\mu_0 - \mu_1)}{S} + z_{\alpha/2}\right)$
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Example: fire protection system

A manufacturer of sprinkler systems used for fire protection in office buildings claims that the true average system-activation temperature is 130 °F. A sample of $n = 90$ systems, when tested, yields a sample average activation temperature of 131.08 °F. If the sample standard deviation of activation times is 2.25 °F **[with NO normal distribution!]**, does the data contradict the manufacturer's claim at significance level $\alpha = 0.05$?

$$H_0 : \text{average activation temperature } \mu = 130 \quad \text{v.s.} \quad H_a : \mu \neq 130$$

- The test: **reject H_0 if $|Z| = \left| \frac{\sqrt{n}(\bar{X}-130)}{S} \right| > z_{0.025} = 1.96$, otherwise accept H_0**
- **Sample z-statistic we observed $z = \frac{\sqrt{9}(131.08-130)}{2.25} = 1.44 < 1.96 \Rightarrow$ accept H_0** , i.e. under 5% significance level, the data does not contradict the manufacturer's claim
- Test statistic $Z = \frac{\sqrt{n}(\bar{X}-\mu_0)}{\sigma} \sim N(0, 1)$
- **p-value = $\mathbb{P}(|Z| > |z|) = \mathbb{P}(|Z| > 1.44) = 2\mathbb{P}(Z > 1.44) = 0.1498 > 0.05 \Rightarrow$ accept H_0**

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H_0 : average activation temperature $\mu = 130$ v.s. H_a : $\mu \neq 130$

- The test: **reject H_0 if $|Z| > 1.96$, otherwise accept H_0**
- When $\mu = 131$ (H_a true), we'd make Type-II error if accept H_0 ($|Z| \leq 1.96$).

Type-II error rate

$$\begin{aligned} &= \mathbb{P}\left(-1.96 \leq \frac{\sqrt{9}(\bar{X} - 130)}{2.25} \leq 1.96\right) \\ &= \mathbb{P}\left(-1.96 - \frac{\sqrt{9}}{2.25} \leq \underbrace{\frac{\sqrt{9}(\bar{X} - 131)}{2.25}}_{\sim N(0,1)} \leq 1.96 - \frac{\sqrt{9}}{2.25}\right) \\ &= \Phi(0.63) - \Phi(-3.29) \approx 0.7352 \end{aligned}$$

Example (revisited): fire protection system

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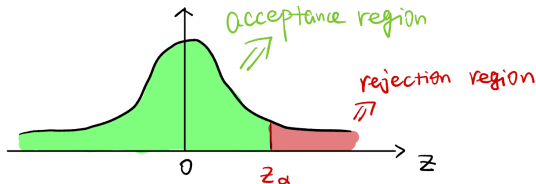
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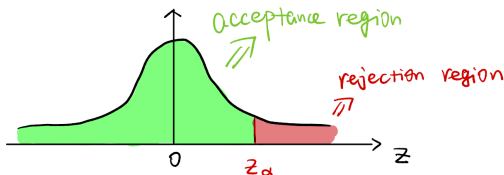
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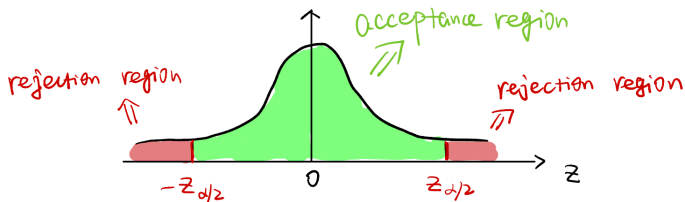
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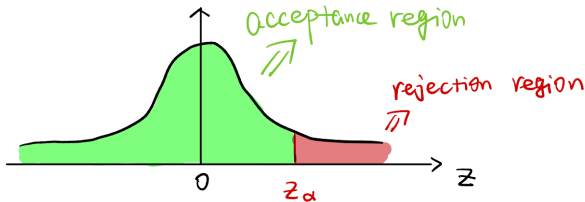


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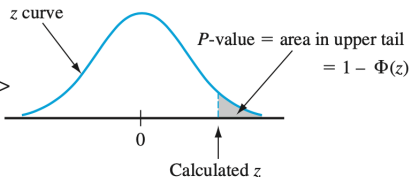


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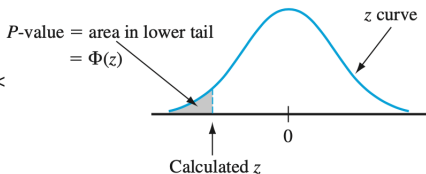


p-value in two-sided and one-sided z-tests

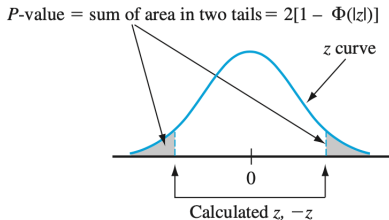
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$$H_0 : \text{average activation temperature } \mu = 130 \quad \text{v.s.} \quad H_a : \mu > 130$$

- The test: **reject H_0 if $Z > z_{0.05} = 1.645$, otherwise accept H_0**
- **Sample z-statistic we observed $z = \frac{\sqrt{9}(131.08-130)}{2.25} = 1.44 < 1.645 \Rightarrow$ reject H_0 , i.e. under 5% significance level, the data contradicts the manufacturer's claim**
- Test statistic $Z = \frac{\sqrt{n}(\bar{X}-\mu_0)}{\sigma} \sim N(0, 1)$
- p-value = $\mathbb{P}(Z > z) = \mathbb{P}(Z > 1.44) = 0.0749 > 0.05 \Rightarrow$ accept H_0

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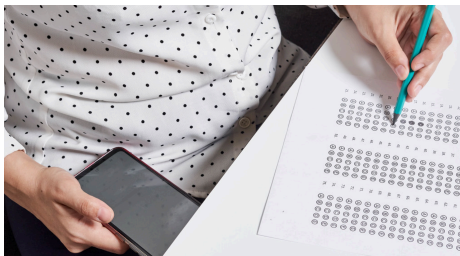
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- **When $\mu = 131$ (H_a true), we'd make Type-II error if accept H_0 ($\bar{X} \leq 130.98$).**

$$\begin{aligned}\text{Type-II error rate} &= \mathbb{P}\left(Z = \frac{\sqrt{9}(\bar{X} - 130)}{2.25} \leq 1.645\right) \\ &= \mathbb{P}\left(\underbrace{\frac{\sqrt{9}(\bar{X} - 131)}{2.25}}_{\sim N(0,1)} \leq 1.645 - \frac{\sqrt{9}}{2.25}\right) \\ &= \Phi(0.31) \approx 0.3783\end{aligned}$$

Large-sample z-test for Population Proportion

Example (new!): proportion of cheating students

27 of the 267 students in a sample admitted to texting during an exam. Can it be concluded at significance level 0.01 that the proportion of students that texted during the exam is significantly different from 5%?



H_0 : the proportion of cheating students $p = 0.05$ v.s. H_a : $p \neq 0.05$

Two-sided large-sample z-test for proportion: derivation

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} \text{Bernoulli}(p)$.

$$H_0 : p = p_0 \quad \text{v.s.} \quad H_a : p \neq p_0,$$

for some constant p_0 . We want to control Type-I error rate $\leq \alpha$.

Step 1: Under H_0 , by CLT, $\bar{p} = \bar{X} \stackrel{d}{\approx} N(p_0, p_0(1-p_0)/n) \Rightarrow$

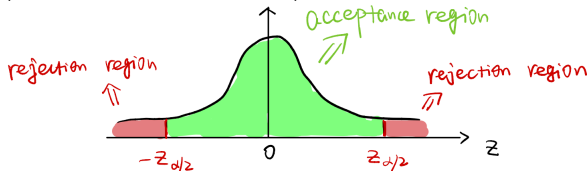
$$Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \stackrel{d}{\approx} N(0, 1)$$

Step 2: The **larger** $|Z|$ is, more likely H_a is true. Therefore the rejection region would be in the form $|Z| > c$. Note that

$$\mathbb{P}(|Z| > z_{\alpha/2}) \approx \alpha.$$

Step 3: The rejection region is $|Z| > z_{\alpha/2}$. p-value = $\mathbb{P}(|Z| > |z|)$

$$= \mathbb{P}\left(|Z| > \left|\frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}}\right|\right) = 2\Phi\left(-\left|\frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}}\right|\right)$$



Two-sided large-sample z-test for proportion

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} \text{Bernoulli}(p)$.

$$H_0 : p = p_0 \quad \text{v.s.} \quad H_a : p \neq p_0,$$

for some constant p_0 . We want to control Type-I error rate $\leq \alpha$.

- Test statistic $Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \stackrel{d}{\approx} N(0, 1)$
- The test: **reject H_0 if $Z < -z_{\alpha/2}$ or $Z > z_{\alpha/2}$, otherwise accept H_0 .**
This is called the **two-sided large-sample z-test for proportion** with level α .
- **p-value** = $\mathbb{P}(|Z| > |z|) = 2\Phi\left(-\left|\frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}}\right|\right)$
- **When $p = p_1$, $\hat{p} \stackrel{d}{\approx} N(p_1, p_1(1-p_1)/n) \Rightarrow \frac{\sqrt{n}(\bar{X}-\mu_1)}{S} \stackrel{d}{\approx} N(0, 1)$.**

$$\text{Type-II error rate} = \mathbb{P}(-z_{\alpha/2} \leq Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \leq z_{\alpha/2}) \approx$$

$$\mathbb{P}\left(\frac{\sqrt{n}(p_0-p_1)}{\sqrt{p_1(1-p_1)}} - \frac{\sqrt{p_0(1-p_0)}}{\sqrt{p_1(1-p_1)}}z_{\alpha/2} \leq \frac{\sqrt{n}(\hat{p}-p_1)}{\sqrt{p_1(1-p_1)}} \leq \frac{\sqrt{n}(p_0-p_1)}{\sqrt{p_1(1-p_1)}} + \frac{\sqrt{p_0(1-p_0)}}{\sqrt{p_1(1-p_1)}}z_{\alpha/2}\right) \\ \approx \Phi\left(\frac{\sqrt{n}(p_0-p_1)}{\sqrt{p_1(1-p_1)}} + \frac{\sqrt{p_0(1-p_0)}}{\sqrt{p_1(1-p_1)}}z_{\alpha/2}\right) - \Phi\left(\frac{\sqrt{n}(p_0-p_1)}{\sqrt{p_1(1-p_1)}} - \frac{\sqrt{p_0(1-p_0)}}{\sqrt{p_1(1-p_1)}}z_{\alpha/2}\right)$$

Example: proportion of cheating students

27 of the 267 students in a sample admitted to texting during an exam. Can it be concluded at significance level 0.01 that the proportion of students that texted during the exam is significantly different from 5%?

H_0 : the proportion of cheating students $p = 0.05$ v.s. H_a : $p \neq 0.05$

- The test: reject H_0 if $|Z| > z_{0.005} = 2.58$, otherwise accept H_0
- Sample proportion $\hat{p} = \frac{27}{267} = 0.10$, sample test statistic we observed $z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} = \frac{\sqrt{267} \times (0.10 - 0.05)}{\sqrt{0.05 \times 0.95}} = 3.75 > 2.58 \Rightarrow$ reject H_0 , i.e. under 5% significance level, there's sufficient evidence that the proportion significantly deviates from 5%
- Under H_0 , test statistic $Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \stackrel{d}{\approx} N(0, 1)$
- p-value = $\mathbb{P}(|Z| > |z|) = \mathbb{P}(|Z| > 3.75) = 2\Phi(-3.75) \approx 0 < 0.01 \Rightarrow$ reject H_0

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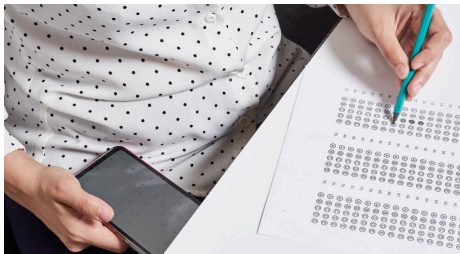
- The test: **reject H_0 if $|Z| > 2.58$, otherwise accept H_0**
- When $p = 0.03$ (H_a true), we'd make Type-II error if accept H_0 ($|Z| \leq 2.58$).

$$\begin{aligned} \text{Type-II error rate} &= \mathbb{P}\left(-2.58 \leq Z = \frac{\sqrt{267}(\hat{p}-0.05)}{\sqrt{0.05 \times 0.95}} \leq 2.58\right) \\ &= \mathbb{P}\left(-2.58 \frac{\sqrt{0.05 \times 0.95}}{\sqrt{0.03 \times 0.97}} + \frac{\sqrt{267} \times 0.02}{\sqrt{0.03 \times 0.97}} \leq \underbrace{\frac{\sqrt{267}(\hat{p} - 0.03)}{\sqrt{0.03 \times 0.97}}}_{\stackrel{d}{\approx} N(0,1)} \leq \right) \end{aligned}$$

$$\begin{aligned} &2.58 \frac{\sqrt{0.05 \times 0.95}}{\sqrt{0.03 \times 0.97}} + \frac{\sqrt{267} \times 0.02}{\sqrt{0.03 \times 0.97}} \\ &\approx \Phi(5.21) - \Phi(-1.38) = 0.9172 \end{aligned}$$

Example (revisited): proportion of cheating students

27 of the 267 students in a sample admitted to texting during an exam. Can it be concluded at significance level 0.01 that **more than** 5% of all students in the population sampled had texted during an exam?



H_0 : the proportion of cheating students $p = 0.05$ v.s. H_a : $p > 0.05$

One-sided large-sample z-test for proportion: derivation

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} \text{Bernoulli}(p)$.

$$H_0 : p = p_0 \quad \text{v.s.} \quad H_a : p > p_0,$$

for some constant p_0 . We want to control Type-I error rate $\leq \alpha$.

Step 1: Under H_0 , by CLT, $\bar{p} = \bar{X} \stackrel{d}{\approx} N(p_0, p_0(1-p_0)/n) \Rightarrow$

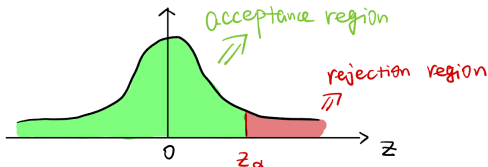
$$Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \stackrel{d}{\approx} N(0, 1)$$

Step 2: The **larger** Z is, more likely H_a is true. Therefore the rejection region would be in the form $Z > c$. Note that

$$\mathbb{P}(Z > z_\alpha) \approx \alpha.$$

Step 3: The rejection region is $Z > z_\alpha$. p-value = $\mathbb{P}(Z > z)$

$$= \mathbb{P}\left(Z > \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}}\right) \approx 1 - \Phi\left(\frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}}\right)$$



One-sided large-sample z-test for proportion

Suppose we have data $X_1, \dots, X_n \stackrel{i.i.d.}{\sim} \text{Bernoulli}(p)$.

$$H_0 : p = p_0 \quad \text{v.s.} \quad H_a : p > p_0,$$

for some constant p_0 . We want to control Type-I error rate $\leq \alpha$.

- Test statistic $Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \stackrel{d}{\approx} N(0, 1)$
- The test: **reject H_0 if $Z < -z_{\alpha/2}$ or $Z > z_{\alpha/2}$, otherwise accept H_0 .**
This is called the **two-sided large-sample z-test for proportion** with level α .
- **p-value** = $\mathbb{P}(Z > z) = 1 - \Phi\left(-\frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}}\right)$
- **When $p = p_1$, $\hat{p} \stackrel{d}{\approx} N(p_1, p_1(1-p_1)/n) \Rightarrow \frac{\sqrt{n}(\bar{X}-\mu_1)}{S} \stackrel{d}{\approx} N(0, 1)$.**

$$\text{Type-II error rate} = \mathbb{P}\left(Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \leq z_{\alpha/2}\right)$$

$$\approx \mathbb{P}\left(\frac{\sqrt{n}(\hat{p}-p_1)}{\sqrt{p_1(1-p_1)}} \leq \frac{\sqrt{n}(p_0-p_1)}{\sqrt{p_1(1-p_1)}} + \frac{\sqrt{p_0(1-p_0)}}{\sqrt{p_1(1-p_1)}} z_{\alpha/2}\right)$$

$$\approx \Phi\left(\frac{\sqrt{n}(p_0-p_1)}{\sqrt{p_1(1-p_1)}} + \frac{\sqrt{p_0(1-p_0)}}{\sqrt{p_1(1-p_1)}} z_{\alpha/2}\right)$$

Example: proportion of cheating students

27 of the 267 students in a sample admitted to texting during an exam. Can it be concluded at significance level 0.01 that **more than** 5% of all students in the population sampled had texted during an exam?

H_0 : the proportion of cheating students $p = 0.05$ v.s. H_a : $p > 0.05$

- The test: **reject H_0 if $Z > z_{0.01} = 2.33$, otherwise accept H_0**
- **Sample proportion $\hat{p} = \frac{27}{267} = 0.10$, sample test statistic we observed $z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} = \frac{\sqrt{267} \times (0.10-0.05)}{\sqrt{0.05 \times 0.95}} = 3.75 > 2.33 \Rightarrow$ reject H_0 , i.e. under 5% significance level, there's sufficient evidence that the proportion significantly deviates from 5%**
- Under H_0 , test statistic $Z = \frac{\sqrt{n}(\hat{p}-p_0)}{\sqrt{p_0(1-p_0)}} \stackrel{d}{\approx} N(0, 1)$
- **p-value = $\mathbb{P}(Z > z) = \mathbb{P}(Z > 3.75) = \Phi(-3.75) \approx 0 < 0.01 \Rightarrow$ reject H_0**

Example: proportion of cheating students

27 of the 267 students in a sample admitted to texting during an exam. Can it be concluded at significance level 0.01 that **more than** 5% of all students in the population sampled had texted during an exam?

H_0 : the proportion of cheating students $p = 0.05$ v.s. H_a : $p > 0.05$

- The test: **reject H_0 if $Z > 2.33$, otherwise accept H_0**
- When $p = 0.08$ (H_a true), we'd make Type-II error if accept H_0 ($Z \leq 2.33$).

$$\begin{aligned}\text{Type-II error rate} &= \mathbb{P}\left(Z = \frac{\sqrt{267}(\hat{p}-0.05)}{\sqrt{0.05 \times 0.95}} \leq 2.33\right) \\ &= \mathbb{P}\left(\underbrace{\frac{\sqrt{267}(\hat{p} - 0.08)}{\sqrt{0.08 \times 0.92}}}_{\stackrel{d}{\approx} N(0,1)} \leq 2.33 \frac{\sqrt{0.05 \times 0.95}}{\sqrt{0.08 \times 0.92}} - \frac{\sqrt{267 \times 0.03}}{\sqrt{0.08 \times 0.92}}\right) \\ &\approx \Phi(0.06) = 0.5239\end{aligned}$$

Many thanks to

- Joyce Robbins
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- And all my teachers in the past 25 years