

SESSIONS 4

Inferential Statistics: Hypothesis Testing Part 3

Data Science Program

Hypothesis test for proportion

- ❑ Z test single population proportion
- ❑ Z test double population proportion
- ❑ Chi-Squared test more than two population proportion

Use Case: A/B Testing categorical measurements

1. An newly launched online store want to test whether their conversion rate is greater than 50% or not. So they **gathered 1000 visitor data randomly** and observed whether they are buying or not. As a Data Scientist you are asked to perform a hypothesis testing whether their conversion rate is greater than 50%?
2. A few months later the online store launched a new web design. They **randomly assign** visitor to layout A or layout B and 1000 for each design. You are asked again to perform hypothesis testing whether the conversion rate is increasing or not ?

Similar Ex. Amazon, Ben, Ubisoft

Z-Test Single Population Proportion

Assumption:

1. Dichotomy categorical variable with population proportion p_0
2. Data collected using randomization
3. sample size (n) large enough:
 $n \times p_0 \geq 15$ and $n \times (1 - p_0) \geq 15$

Hypothesis:

$H_0 : p = p_0$ (ex. $p = 0.5$)

$H_0 : p \neq p_0$ (two sided) or

$p < p_0$ or $p > p_0$ (one sided)

Test Statistics z:

$$z = \frac{\hat{p} - p_0}{se_0} = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1 - p_0)}{n}}}$$

Rejection Criteria (spicy) :

P-value $\leq \alpha$ (two-sided)

P-value/2 $\leq \alpha$ (one-sided)

Z-Test Double Population Proportion

Assumption:

1. Dichotomy categorical variable observed in two group
2. Data collected using randomization
3. sample size (n_1 and n_2) large enough that there are five successive and five failure in each group.

Hypothesis:

$H_0 : p_1 = p_2$ (ex. $p_1 - p_2 = 0$)

$H_0 : p \neq p_0$ (two sided) or

$p_1 < p_2$ or $p_1 > p_2$ (one sided)

Test Statistics z:

$$z = \frac{(\hat{p}_1 - \hat{p}_2) - 0}{se_0} \text{ with } se_0 = \sqrt{\hat{p}(1 - \hat{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}$$

where \hat{p} is the pooled estimate.

Rejection Criteria (spicy) :

$P\text{-value} \leq \alpha$ (two-sided)

$P\text{-value}/2 \leq \alpha$ (one-sided)

Use Case: A/B Testing categorical measurements more than two groups

- A few months later the online store launched a new web design. They **randomly assign** visitor to layout A (current design), layout B1, or layout B2 and 1000 for each design.
- You are asked again to perform hypothesis testing whether the conversion rate is increasing or not and which design perform best?

Similar Ex. Zalora Case

Chi-Squared Test for Independence between Two Categorical Variables

Assumption:

1. Two categorical variable :
check out (yes or no) vs web design (design A, B1, B2)
2. Randomization

Hypothesis:

H_0 : The two variables are independent
 H_a : The two variables are dependent.

Test Statistics χ^2 :

$$\chi^2 = \sum \frac{(\text{observed count} - \text{expected count})^2}{\text{expected count}}$$

Rejection Criteria: $P\text{-value} \leq \alpha$

A/B Testing part 2

- Logistics Regression should be used in A/B Testing for Zalora use case.
- A/B Testing with more advanced method (ex using logistics regression) should be discuss in module 3 after learning regression and logistics regression.

Goodness of Fit Test

- ❑ We need normality assumption for some test
- ❑ We can check the distribution but that is very vulnerable to subjectivity
- ❑ To Test if sample data follow certain distribution (ex. Normal)
- ❑ Commonly used goodness of fit test:
 - Chi - square
 - Anderson - darling
 - Shapiro - wilk
 - Kolmogorov – smirnov

Hypothesis

Ho : The sample follow certain distribution

H1 : The sample do not follow the distribution

Rejection Region : $P\text{-Value} < \alpha$

Non-Parametric

Non-parametric Statistics usually used when

- We don't have enough data (small sample size)
- We can't assume that our data population follow certain distribution. ex. result of our normality test is reject H_0

	Parametric	Non-parametric
Assumed distribution	Normal	Any
Assumed variance	Homogeneous	Any
Data set relationships	Independent	Any
Usual central measure	Mean	Median
Benefits	Can draw more conclusions	Simplicity; Less affected by outliers
Tests		
Correlation test	Pearson	Spearman
Independent measures, 2 groups	Independent-measures t-test	Mann-Whitney test
Independent measures, >2 groups	One-way, independent-measures ANOVA	Kruskal-Wallis test
Repeated measures, 2 conditions	Matched-pair t-test	Wilcoxon test
Repeated measures, >2 conditions	One-way, repeated measures ANOVA	Friedman's test

Hypothesis testing for mean: Non-Parametrics

- ❑ **Mann Whitney** for independence double population mean
 - ❑ **Wilcoxon** for paired mean test
 - ❑ **Kruskal Wallis** for more than two population mean
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- ✓ These methods are usually used when normality assumption doesn't meet
 - ✓ For example: Income data or money related data usually have right skew distribution

Wilcoxon

Assumption:

1. numerical value with population mean **μ_0**
2. Data collected using randomization
3. Test for median of population
4. sample size too small

Hypothesis:

$H_0 : \mu = \mu_0$ (ex. $\mu = 40$)

$H_0 : \mu \neq \mu_0$ (two sided) or

$\mu < \mu_0$ or $\mu > \mu_0$ (one sided)

Test Statistics : Sumrank

Rejection Criteria (spicy) :

$P\text{-value} \leq \alpha$ (two-sided)

$P\text{-value}/2 \leq \alpha$ (one-sided)

Mann Whitney

Assumption:

1. numerical variable (ordinal, interval, ratio) in each group
2. Data collected using randomization
3. Test for median of population
4. sample size too small

Hypothesis:

$H_0 : M1 = M2$ (ex. $M1 - M2 = 0$)

$H_0 : M1 \neq M2$ (two sided) or

$M1 < M2$ or $M1 > M2$ (one sided)

Test Statistics : U-Statistics

Rejection Criteria (spicy) :

$P\text{-value} \leq \alpha$ (two-sided)

$P\text{-value}/2 \leq \alpha$ (one-sided)

Kruskal-Wallis

Assumption:

1. The population distribution of the numerical variable is assumed to be normal
2. Data collected using randomization
3. Test for median of population
4. sample size too small

Hypothesis:

$H_0 : M_1 = M_2 = \dots = M_k$ (there are k groups)

H_a : at least one pair of the population are not equal (and we don't know which)

Test Statistics: V-Statistics

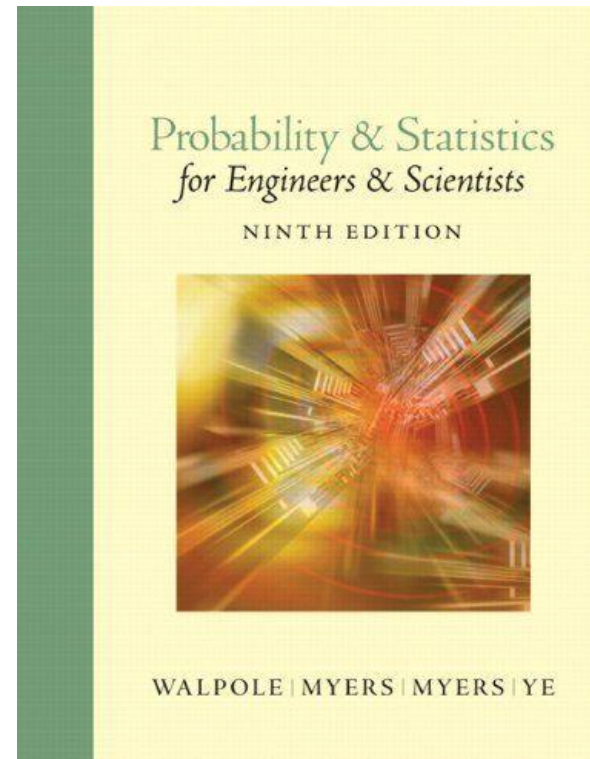
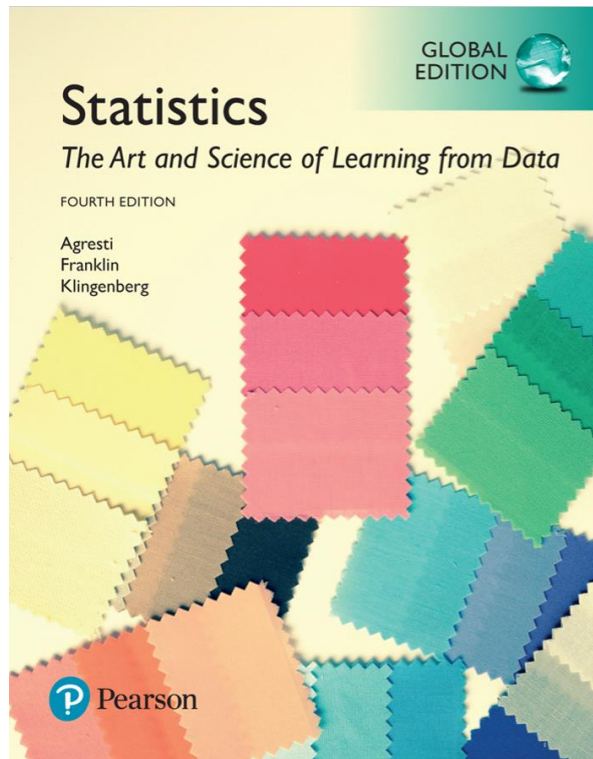
Rejection Region: $p\text{-value} < \alpha$

Another example of Hypothesis testing

Methods that you have learned can also be applied to another cases:

- Medicine, to understand if a drug works or not
- Economics, to understand human behavior
- Foreign aid and charitable work (the reputable ones at least), to understand which interventions are most effective at alleviating problems (health, poverty, etc)

Reference



Reference

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