

29 三 . 16 次课

1. lecture
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5. 研究



Panacea
万金油
万灵丹
||
Fisher Randomization Test
(FRT)

The Design of Experiments

By

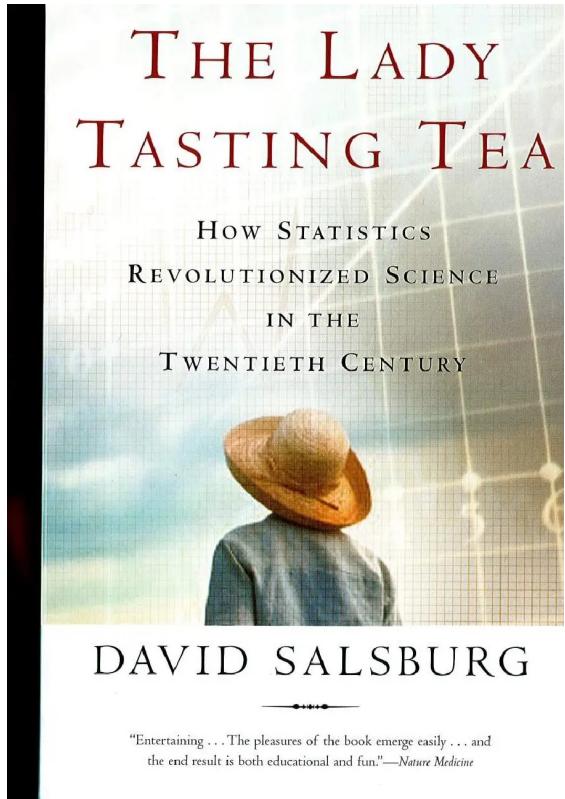
Sir Ronald A. Fisher, Sc.D., F.R.S.

Honorary Research Fellow, Division of Mathematical Statistics, C.S.I.R.O., University of Adelaide; Foreign Associate, United States National Academy of Sciences; and Foreign Honorary Member, American Academy of Arts and Sciences; Foreign Member of the Swedish Royal Academy of Sciences, and the Royal Danish Academy of Sciences and Letters; Member of the Pontifical Academy; Member of the German Academy of Sciences (Leopoldina); formerly Galton Professor, University of London, and Arthur Balfour Professor of Genetics, University of Cambridge.

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Fisher(1935)

First Published 1935
Second Edition 1937
Third Edition 1942
Fourth Edition 1947
Fifth Edition 1949
Sixth Edition 1951
Reprinted 1953
Seventh Edition 1960
Eighth Edition 1966
Ninth Edition 1971
Reprinted 1974



"Entertaining . . . The pleasures of the book emerge easily . . . and the end result is both educational and fun."—Nature Medicine

女士品茶

书 3.5.2

lady milk lady tea

Fisher milk first	X	4-X	4
Fisher tea first	4-X	X	4
	4	4	

$$P_{FRT} = \frac{1}{\binom{8}{4}} = \frac{1}{70} = 0.014 < 0.05$$

 CAPITAL OF STATISTICS
PROFESSION, HUMANITY & INTEGRITY

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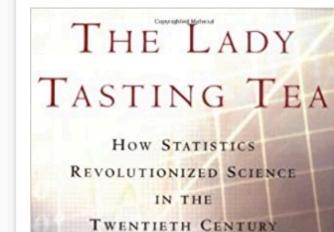
女士品茶的实验、假设和检验

丁鹏

关键词: 实验设计; 假设检验

编辑: 于森、林枫 审稿: 黄湘云、魏太云

R. A. Fisher 的名著《实验设计; 第八版, 1971 年》第二部分有十六页, 仅仅讲了一个最简单的实验: 女士品茶。这个故事非常有名, 以至于 Salsburg 的统计学通俗读物就以它命名: 《女士品茶: 20 世纪统计怎样变革了科学》。

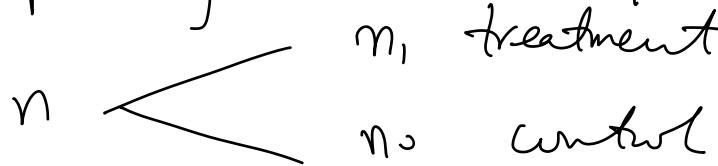


Chapter 3

CRE

FRT

CRE : completely randomized experiment

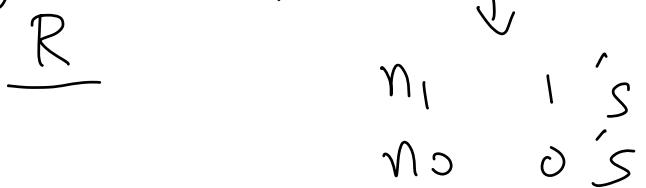


n, n_1, n_0 fixed

$$\vec{z} = (z_1, z_2, \dots, z_n)$$

$$Pr(\vec{z} = \vec{z}) = \frac{1}{\binom{n}{n_1}}$$

computationally : sample (\vec{z})



$$Pr(\vec{z} | \vec{Y}(1) - \vec{Y}(0)) = Pr(\vec{z})$$

= uniform

treatment assignment mechanism

~~Stages of the function~~

Rubin's
terminology

$$H_0 F : Y_{i(1)} = Y_{i(0)} \quad \forall i=1 \dots n$$

sharp null

strong null

$$\Rightarrow Y_{i(1)} = Y_{i(0)} = Y_i = Z_i Y_{i(1)} + (-Z_i) Y_{i(0)}$$

are all fixed

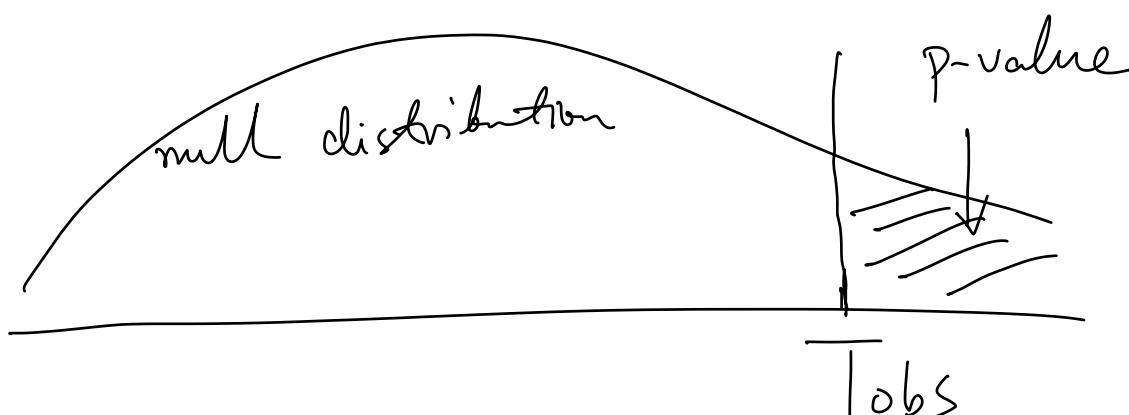
Data	i	Z	Y	
	1	1	✓	fixed
	2	:	✓	
	:	1	✓	
	n	0	✓	
		0	✓	
		0	✓	

under $H_0 F$

random

Test statistic: $T = T(\vec{Z}, \vec{Y})$

measures deviation from $H_0 F$



$$\vec{z} \sim \text{Uniform}(\vec{z}^1, \dots, \vec{z}^M)$$

$$M = \binom{n}{n_1}$$

$$\Rightarrow T(\vec{z}, \vec{y}) \sim \text{Uniform}(T(\vec{z}^1, \vec{y}), \dots, T(\vec{z}^M, \vec{y}))$$

$$P_{FRT} = \frac{1}{M} \sum_{m=1}^M 1 \left\{ T(\vec{z}^m, \vec{y}) \geq T(\vec{z}, \vec{y}) \right\}$$

SS

$$\hat{P}_{FRT} = \frac{1}{R} \sum_{r=1}^R 1 \left\{ T(\vec{z}^r, \vec{y}) \geq T(\vec{z}, \vec{y}) \right\}$$

$R \sqrt{\frac{1}{R}}$ 隨機重複
sample (\vec{z})

$$\text{hw: } \text{Var}(\hat{P}_{FRT}) \leq \frac{1}{4R}$$

T 的选择:

$$\text{eg. } \hat{\tau} = \frac{\hat{\mu}_1 - \hat{\mu}_0}{\sqrt{\frac{1}{n_1} \sum_{z_i=1}^1 Y_i} + \sqrt{\frac{1}{n_0} \sum_{z_i=0}^0 Y_i}}$$

$$\text{eg. 2 } t_{unequal} = \frac{\hat{\mu}_1 - \hat{\mu}_0}{\sqrt{\frac{\hat{s}^2_{(1)}}{n_1}} + \sqrt{\frac{\hat{s}^2_{(0)}}{n_0}}}$$

样本方差

$$\frac{1}{n_1-1} \sum_{z_i=1}^1 \left(Y_i - \hat{\mu}_1 \right)^2$$

R: t.test(.)

var.equal = FALSE

古典统计: Behrens-Fisher 问题

推荐: 不用 FRT 检验 $\left\{ \begin{array}{l} H_0 F: \bar{\tau}_1 = 0 \\ H_0 N: \tau = 0 \end{array} \right.$

那么 FRT + tUnequal 更强
理论保证

e.g 3 Wilcoxon 積分法

$$\text{pool}(Y_1 \dots Y_n)$$

$$\Rightarrow \text{rank}(Y_1 \dots Y_n)$$

積分 ranks: $R_1 \dots R_n$

$$R_i = \#\left\{j : Y_j \leq Y_i\right\}$$

$$W = \sum_{i=1}^n z_i R_i$$

$$\Leftrightarrow \hat{R}(1) - \hat{R}(0)$$

e.g 4 Kolmogorov-Smirnov 檢定

$$\hat{F}_1(y) = \frac{1}{n_1} \sum_{i=1}^{n_1} z_i \mathbf{1}(Y_i \leq y)$$

$$\hat{F}_0(y) = \frac{1}{n_0} \sum_{i=1}^{n_0} (1-z_i) \mathbf{1}(Y_i \leq y)$$

$$D = \max_y \left| \hat{F}_1(y) - \hat{F}_0(y) \right|$$

eg 5 一般策略

① Neyman - Pearson : $H_0 \leftrightarrow H_1$
 $f_0(y)$ $f_1(y)$

$$LRT = \frac{f_1(y)}{f_0(y)}$$

功效
power

② model-assisted

$$\text{lm}(Y_i \sim Z_i) \Rightarrow \hat{\tau} = \hat{Y}_{(1)} - \hat{Y}_{(0)}$$

$$\text{lm}(Y_i \sim Z_i + X_i) \quad \begin{array}{l} \text{协变量} \\ \downarrow \\ \text{covariates} \end{array}$$

$\hat{\tau}_F$: Fisher's ANCOVA

$$\text{glm}(Y_i \sim Z_i)$$

$$\text{glm}(Y_i \sim Z_i + X_i)$$

檢定 FRT: 固定

$$H_0(\vec{\tau}): Y_i(1) - Y_i(0) = \bar{\tau}_i, \quad \forall i$$

$$+ (\bar{z}_i, Y_i)_{i=1}^n,$$

$$\Rightarrow (\bar{z}_i, Y_i(1), Y_i(0))_{i=1}^n$$

計算 $P_{FRT}(\vec{\tau})$

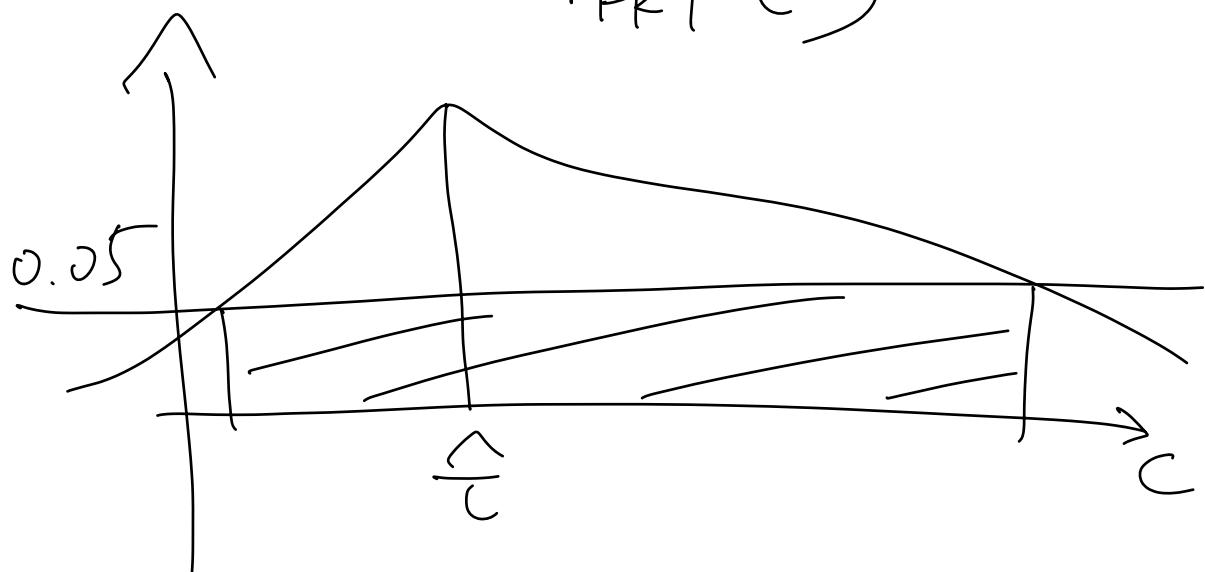
檢定
統計量 $\left\{ \bar{\tau} = \frac{1}{n} \sum_{i=1}^n \bar{\tau}_i : P_{FRT}(\vec{\tau}) > \alpha \right\}$

檢定 $\bar{\tau} \sim (1-\alpha)$ 分位數

的置信區間.

Li & Ding (2016) : $Y = \text{值}$
 $O(3^n) \rightarrow O(n^2)$

Rosenbaum: $H_0(c): Y_{i(1)} - Y_{0(0)} = c$
 假設 c 為 \hat{c}
 $\hat{P}_{FRT}(c)$



testable implication:

$$\hat{F}_1(y) \neq \hat{F}_0(y)$$

Ding Feller Miratrix (2014 JRSSB)

Chapter 4

Neyman (1923)

$$(Y_{i(1)}, Y_{i(0)}) \stackrel{i=1}{\stackrel{n}{\sim}} \Rightarrow \tau = \frac{1}{n} \sum_{i=1}^n (Y_{i(1)} - Y_{i(0)}) \\ = \bar{Y}_{(1)} - \bar{Y}_{(0)}$$

CRE: $\bar{z} \sim \text{Uniform}$

解釋 $Y_i = z_i Y_{i(1)} + (1-z_i) Y_{i(0)}$

无偏估计 $\hat{\tau}$: $\hat{\tau} = \frac{1}{n} \bar{Y}_{(1)} - \frac{1}{n} \bar{Y}_{(0)}$

$$\bar{Y}_{(1)} = \frac{1}{n} \sum_{i=1}^n Y_{i(1)}, \quad \bar{Y}_{(0)} = \frac{1}{n} \sum_{i=1}^n Y_{i(0)}$$

$$\frac{1}{n} \bar{Y}_{(1)} = \frac{1}{n_1} \sum_{z_i=1} Y_i, \quad \frac{1}{n} \bar{Y}_{(0)} = \frac{1}{n_0} \sum_{z_i=0} Y_i$$

$\mathbb{E} \left(\frac{\hat{\tau}}{\tau} \right) = 1$

↓ 意思?

$Y_{i(1)}, Y_{i(0)}$ 同樣

z_i 隨機

$$\hat{\tau} = \frac{1}{n_1} \sum_{i=1}^{n_1} z_i Y_{i(1)} - \frac{1}{n_0} \sum_{i=1}^{n_0} (-z_i) Y_{i(0)}$$

$$= \frac{1}{n_1} \sum_{i=1}^{n_1} z_i Y_{i(1)} - \frac{1}{n_0} \sum_{i=1}^{n_0} (-z_i) Y_{i(0)}$$

$$\Rightarrow E\left(\hat{\tau}\right) = \frac{1}{n_1} \underbrace{\sum_{i=1}^{n_1} E(z_i) Y_{i(1)}}_{= n_1} - \frac{1}{n_0} \sum_{i=1}^{n_0} (-E(z_i)) Y_{i(0)}$$

$$= \frac{1}{n} \sum_{i=1}^n Y_{i(1)} - \frac{1}{n} \sum_{i=1}^n Y_{i(0)} = \bar{\tau}$$

T.R. iGnA

$$\text{方差: } \text{Var}(\hat{\tau}) = \frac{S_{(1)}^2}{n_1} + \frac{S_{(0)}^2}{n_0} - \frac{S(\bar{\tau})^2}{n}$$

其

$$S_{(1)}^2 = \frac{1}{n_1 - 1} \sum_{i=1}^{n_1} \left(\bar{Y}_{i(1)} - \bar{\bar{Y}}_{(1)} \right)^2$$

$$S_{(0)}^2 = \frac{1}{n_0 - 1} \sum_{i=1}^{n_0} \left(\bar{Y}_{i(0)} - \bar{\bar{Y}}_{(0)} \right)^2$$

回顧: $Y_i | z_i = 1 \sim \text{iid} [M_1, \sigma_1^2]$

$Y_i | z_i = 0 \sim \text{iid} [M_0, \sigma_0^2]$

$$\hat{\tau} = \bar{Y}_{(1)} - \bar{Y}_{(0)} \sim \left[M_1 - M_0, \frac{\sigma_1^2}{n_1} + \frac{\sigma_0^2}{n_0} \right]$$

方差估计: $\hat{V} = \frac{\hat{S}^2(1)}{n_1} + \frac{\hat{S}^2(0)}{n_0}$

样本方差

偏序: $E(\hat{V}) - \text{var}(\hat{V}) = \frac{\hat{S}^2(2)}{n} \geq 0$

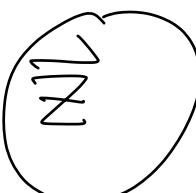
无序

不偏序: $T_i = \bar{T} \quad \forall i$

区间估计: $\bar{T} \pm 1.96 \sqrt{\hat{V}}$

置信

$\left\{ \begin{array}{l} ① \quad \bar{T} \approx N(\bar{T}, \text{var}(\bar{T})) \\ ② \quad \hat{V} \approx \frac{\hat{S}^2(1)}{n_1} + \frac{\hat{S}^2(0)}{n_0} \end{array} \right.$

随机性 \sim 

Hoeffding's combinatorial CLT

同様文献 $A = (a_{ij})$

$$\sum_{i=1}^n a_{i\pi(i)}$$

$$\pi: \{1 \dots n\} \rightarrow \{1 \dots n\}$$

$$\approx \text{Normal}$$

Neyman's final results:

$$\frac{\Delta}{T}, \quad \frac{\nabla}{V}$$