

Date: Tuesday, April 2, 2019

Example of jelly beans causing acne:

- 20 colors investigated
- $\alpha = 0.05$

H_{0j} : Color j jelly bean does not lead to acne
 H_{Aj} : It does

$P(\text{reject at least one null hypothesis} \mid \text{all were true})$

$$= P(FD \geq 1)$$

\hookrightarrow False discovery

FWER

$$= P\left(\bigcup_{i=1}^{20} \{\text{reject } H_{0j} \mid H_{0j} \text{ true}\}\right)$$

(Boole's
ineq.)

$$\leq \sum_{j=1}^{20} P(\text{reject } H_{0j} \mid H_{0j} \text{ true})$$

$$= \sum_{j=1}^{20} 0.05 = 20(0.05) = 1.$$

\hookrightarrow Without controlling for FWER, we can expect (with high probability) to reject at least one H_{0j}
 \hookrightarrow incorrectly
 \Rightarrow we'll find that some color jelly bean causes acne.

$$P(FD \geq 1) = 1 - P(FD = 0) = 1 - (1 - 0.05)^{20}$$

(because of indep.)

$$= 1 - (0.95)^{20}$$
$$= 0.64$$

* So we should correct for multiple comparisons!

First strategy: Control for FWER.

Method: Bonferroni's correction.

- Reject H_0j when the p-value of the test $p_j \leq \frac{\alpha}{m}$

This will guarantee that $\text{FWER} \leq \alpha$!

Jelly bean example:

$$\begin{aligned}\text{FWER} &= 1 - \left(1 - \frac{0.05}{20}\right)^{20} \\ &= 0.0488 \quad \checkmark\end{aligned}$$

But now our p-value cut-off is drastically reduced

⇒ leads to rarely rejecting

⇒ leads to increased type II error for each test

⇒ leads to low power for each test.

Typically, we refer to Bonferroni as a conservative strategy because we will have few rejections.

* There are applications where being conservative on rejection is desired.

Ex: Malware detection
Pharmaceutical trials
Gene activation

Question: How can we specify a multiple comparisons strategy that is less conservative? (and more powerful?)

- 1) We can try alternative methods to control FWER, but in general these are still conservative.
- 2) We can look at alternative criteria to FWER. A popular one is the false discovery rate, **FDR**.

Recall: FWER — $P(\text{having at least one FD in } m \text{ tests})$

↳ To slacken the restriction of FWER, we can allow for a few more FD's.

⇒ FDR does this!

Recall: $R = \#$ of rejections from M tests
 $V = \#$ of false rejections.

$\frac{V}{R} =$ The rate of false discovery but note that V is an unknown random quantity.

$$\Rightarrow \text{FDR} = E \left[\frac{V}{R} \mid R \geq 1 \right] = \frac{E[V \mid R \geq 1]}{\max\{R, 1\}}$$

We'd like to control FDR!

i.e. to ensure that $\text{FDR} \leq \alpha$. (1)

↳ Thankfully, in 1995 Benjamin and Hochberg developed a strategy that guarantees (1)

↳ The paper "Controlling the false discovery rate:..." is the **most** cited statistical paper out there (53,710 as of April 2nd, 2019) compared to the Lasso paper (from 1996 w/27549).

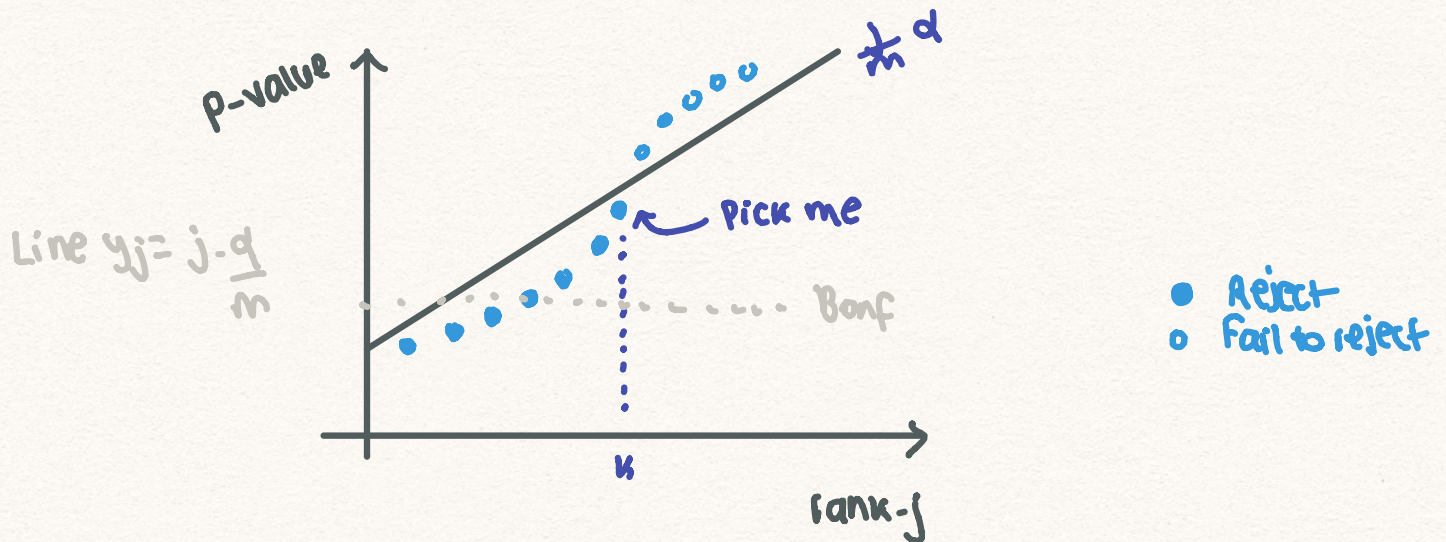
B-H step-up procedure

Setup: 1) Orders p-values from smallest to largest

$$P_{(1)} \leq P_{(2)} \leq \dots \leq P_{(m)} \quad \text{where } P_{(j)} = j^{\text{th}} \text{ smallest p-value}$$

2) Find the largest k such that

$$P_{(k)} \leq \frac{k}{m} \alpha \quad \text{and} \quad P_{(j)} \leq \frac{j}{m} \alpha \quad \text{for all } j=1, \dots, k$$



3) Reject all hypothesis $j = 1, \dots, k$

Notes:

• As the rank of the p-value increases, the cutoff increases linearly.

- Ex:
- $P_{(1)}$ is rejected if $\leq \frac{\alpha}{m}$ (Bonferroni!)
 - $P_{(m)}$ is rejected if $\leq \alpha$ (no adjustment!)

Key Property: If the hypothesis tests are independent, then running the set-up procedure guarantees FDR $\leq \alpha$.

If not independent, this holds approx. as $m \rightarrow \infty$

In class - exercise:

- Simulate 1,000 $U(0,1)$ \rightarrow treat as p-values
 - 1) Reject w/ no corr
 - 2) Bonferroni
 - 3) B-H
 - * rejection)

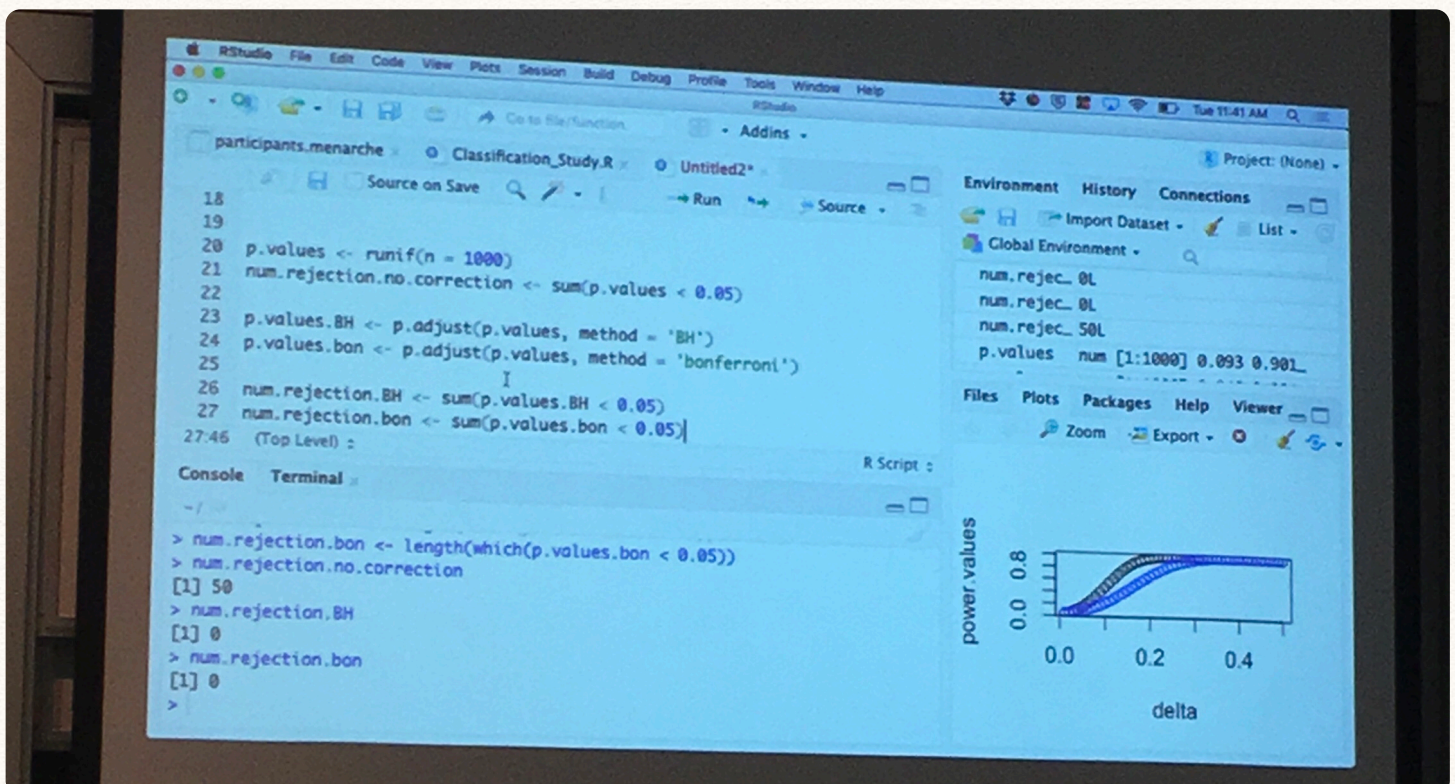
Note: Can approximate $FDR = \frac{E[V | R \geq 1]}{\max\{R, 1\}}$
using Monte Carlo simulation ☺

* Once we run step-up (or Bonferroni) we can calculate the observed false discovery proportion (fdr) as $fdr = \frac{v}{R}$ \leftarrow observed if we know which hypoth are true.

* If H_0 is true, one can show that $p\text{-val} \sim U(0,1)$

* Bonferroni, a corrected p-value will be $m \cdot p\text{-value}$ (if $\leq \alpha$, reject)

* In step up, it will be for the i th smallest p-value $p_{(i)}^{(row)} = \frac{m \cdot p_{(i)}}{i}$ (if $\leq \alpha$, reject)



* How do FDR & FWER compare?

1) If all H_{0j} are true, $FDR = FWER$

2) In general, $FDR \leq FWER \leq \alpha$

\Rightarrow If we control FWER, we also control FDR.

$$FDR \approx P(V > 1) \leq P(V \geq 1)$$