

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

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

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

↳ False discovery

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

$$\stackrel{\text{(Boole's incr.)}}{\leq} \sum_{j=1}^{20} P(\text{reject } H_{0j} \mid H_{0j} \text{ true})$$

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

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

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

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

$$0.64$$

(Because
of indep.)

* So we should correct for multiple comparisons!

First strategy: Control for FWER.

Method: Bonferroni's correction.

- Reject H_0 when the p-value of the test $p_j \leq \frac{\alpha}{m}$

This will guarantee that $FWER \leq \alpha$!

Jelly bean example:

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

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 criterions 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 FPs.

⇒ FDR does this!

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

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

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

We'd like to control FDR!

i.e. to ensure that $FDR \leq \alpha$. (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 April 2nd, 2019) compared to the Lasso paper (from 1996 w/ 27549).

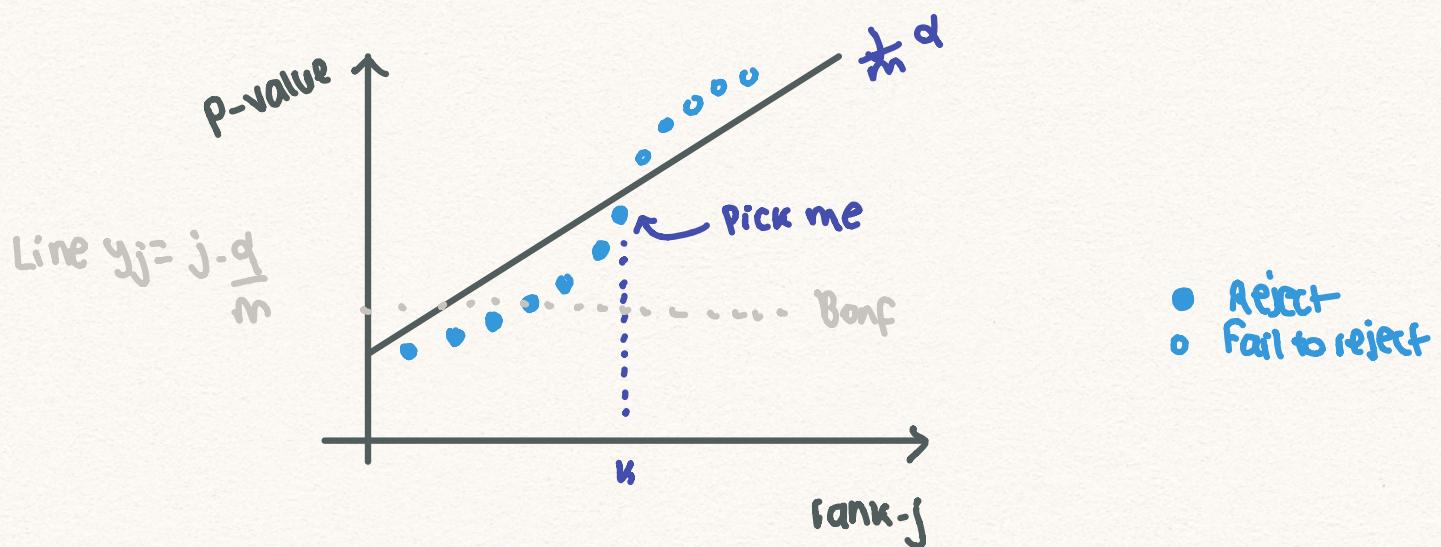
B-H step-up Procedure

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 $m \rightarrow \infty$.

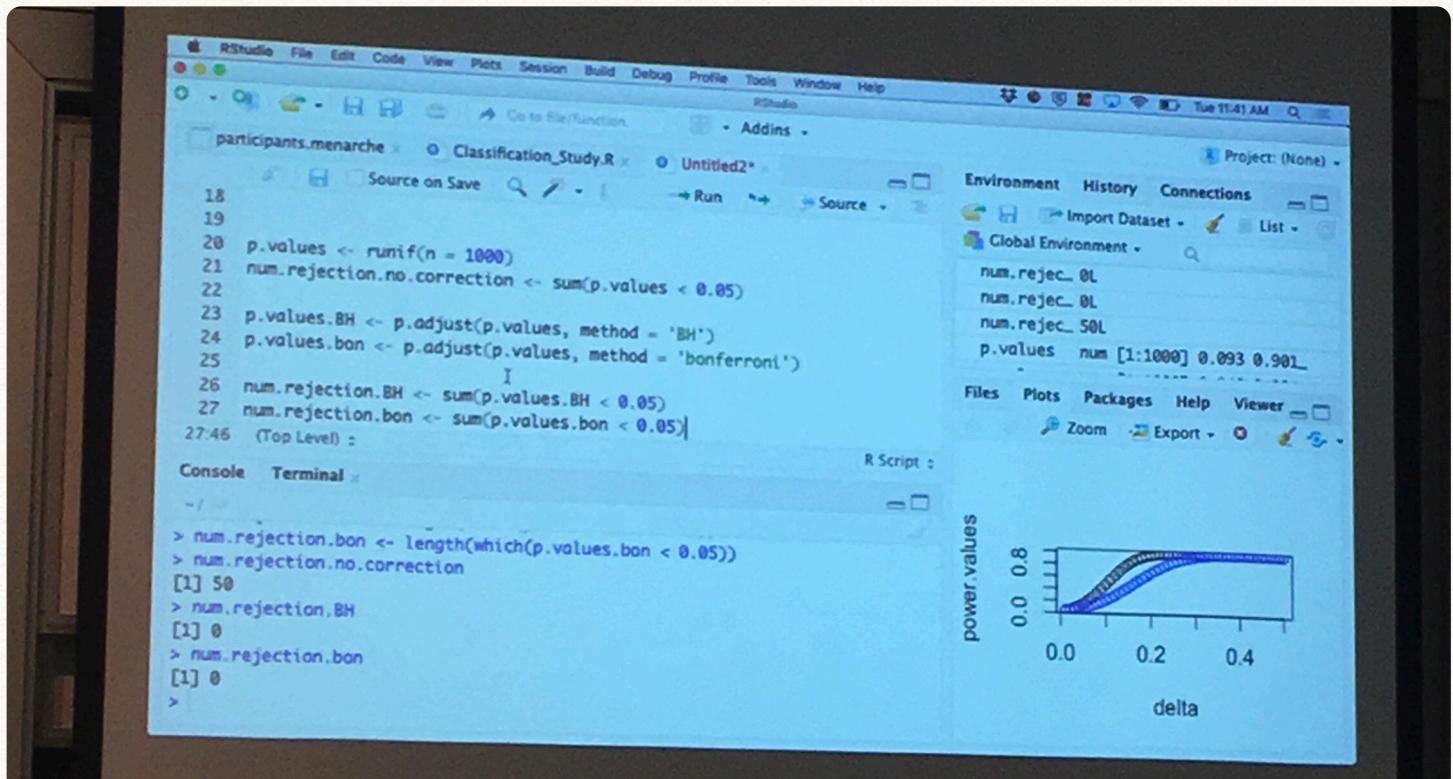
IN CLASS - EXERCISE:

- 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 (fdp) as $fdp = \frac{X}{n} \leftarrow$ observed if we know which hypoth are true.
- * If H_0 is true, one can show that p-val $\sim U(0,1)$
- * Bonferroni, a collected p-value will be m. p-value (if $\leq \alpha$, reject)
- * In step up, it will be for the i-th smallest p-value $p_{(i)}^{(new)} = \frac{m-p(i)}{i}$ (if $\leq \alpha$, reject)



- * How do FDR & FWER compare?
 - 1) If all H_0j are true, $FDR = FWER$
 - 2) In general, $FDR \leq FWER \leq \alpha$
⇒ If we control FWER, we also control FDR.
 $FDR \approx P(\vee \gamma_i) \leq P(\vee \gamma_i)$