

$$\text{Prof: (i) } \text{FDR}(R^{\text{BH}}, P) = \sum_{j=1}^m (1-\theta_j) \mathbb{E}\left[\frac{\mathbf{1}\{P_j \leq \alpha(\hat{e}_{V1})/m\}}{\hat{e}_{V1}}\right]$$

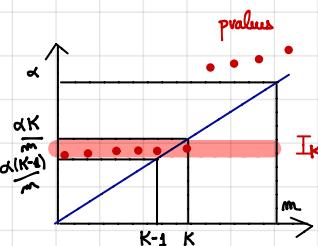
$$\text{Use that for any } l \in \{1, \dots, m\}, \frac{1}{l} = \sum_{k \geq l} \left( \frac{1}{k} - \frac{1}{k+1} \right) = \sum_{k \geq l} \frac{1}{k(k+1)}$$

$$\text{Therefore } \mathbb{E}\left[\frac{\mathbf{1}\{P_j \leq \alpha(\hat{e}_{V1})/m\}}{\hat{e}_{V1}}\right] = \mathbb{E}\left[\sum_{k=1}^m \frac{1}{k(k+1)} \mathbf{1}\{P_j \leq \alpha(\hat{e}_{V1})/m\} \mathbf{1}\{k \geq l\}\right]$$

$$\leq \sum_{k \geq l} \frac{1}{k(k+1)} \mathbb{P}(P_j \leq \frac{\alpha(k)m}{m}, k \geq \hat{e}_{V1}) \leq \frac{\alpha}{m} \sum_{k \geq l} \frac{k \cdot m}{k(k+1)} \text{ because } \theta_j = 0$$

$$\sum_{k \geq l} \frac{k \cdot m}{k(k+1)} = \sum_{k=1}^m \frac{1}{k+1} + \sum_{k > m} m \cdot \frac{1}{k(k+1)} = 1 + \frac{1}{2} + \dots + \frac{1}{m}.$$

(ii) assume  $\alpha \gamma_m \leq 1$ . We build a special distribution on  $[0, 1]^m$  in the following way



$$\text{Let } I_k = \left(\frac{\alpha(k-1)}{m}, \frac{\alpha k}{m}\right], 1 \leq k \leq m \text{ and } I_0 = (\alpha, 1]$$

\*  $K$  is drawn on  $\{0, 1, \dots, m\}$  with  $P(K=k) = \frac{\alpha}{k}$ ,  $1 \leq k \leq m$   
 $P(K=0) = 1 - \alpha \gamma_m$

\* conditionally on  $\{K=k\}$ , pick  $S_k$  uniformly in the subsets  $\{1, \dots, m\}$  of cardinal  $k$

- pick  $U_j, j \in S_k$ , iid uniform in  $I_k$
- pick  $U_j, j \notin S_k$  iid uniform in  $I_0$

Then each  $U_j$  has a uniform distribution:

$$\forall j \in \{1, \dots, m\}, \forall k \in \{1, \dots, m\}, P(U_j \in I_k) = P(U_j \in I_k, K=k) = P(j \in S_k, K=k)$$

$$= P(j \in S_k | K=k) \times P(K=k)$$

$$= \frac{k}{m} \times \frac{\alpha}{k} = \frac{\alpha}{m}$$

and, clearly,  $U_j$  uniform on  $I_k$  cond. on "  $U_j \in I_k$ "

Now, by assumption,  $\exists P_0 \in \mathcal{P}$  such that  $\Theta(P_0) = \emptyset$  (full null) and

$$(P_j | X), 1 \leq j \leq m \sim N(X P_0, I_m)$$

We merely check that the number of rejections of BH procedure satisfies  $\hat{e}_N K$

$$\text{Therefore } \text{FDR}(R^{\text{BH}}, P_0) = P(\hat{e} > 1) = P(K \geq 1) = \alpha \gamma_m$$

If  $\alpha \gamma_m > 1$  i.e.  $\alpha > \gamma_m^{-1} := \alpha^*$  choose  $P_0$  as before with  $\alpha^*$  instead of  $\alpha$  (so that  $\alpha^* \gamma_m = 1$ )

$$\text{FDR}(R^{BH(\alpha)}, P_0) = P(|R^{BH(\alpha)}| \geq 1) \geq P(|R^{BH(\alpha^*)}| \geq 1) = \text{FDR}(R^{BH(\alpha^*)}, P_0) = 1 \quad \square$$

↑ from above

Consequence:  $R^{BH(\frac{\alpha}{\gamma_m})}$  always controls the FDR at level  $\alpha$ . called the BY procedure

**Exercise** generalization of BY procedure [Blanchard and R. (2008)]

Consider  $p(e) = \sum_{k=1}^e k \gamma_k$  with  $\{\gamma_k, 1 \leq k \leq m\}$  some pmba measure

Show that the step-up stopping rule  $\hat{e} = \max \{e \in \{0, \dots, m\} : p(e) \leq \alpha \frac{p(e)}{m}\}$  always provides a control at level  $\alpha$ .

But: known to be conservative procedure for a 'realistic'  $P$   
so we prefer BH

## ② Dependence conditions ensuring that BH control the FDR

**Theorem:** consider p-values satisfying (\*) and the BH procedure at level  $\alpha$  with  $\hat{e}$  rejections  
Assume that  $P \in \mathcal{P}$  is such that

$$\forall j \text{ with } \theta_j = 0, \forall e \in \{2, \dots, m\}, P(\hat{e} \leq e-1 \mid p_j \leq \frac{\alpha(p-e)}{m}) \leq P(\hat{e} \leq e-1 \mid p_j \leq \frac{\alpha e}{m}) \quad (\#)$$

Then  $\text{FDR}(R^{BH}, P) \leq \frac{m_0(P)}{m} \alpha$

**Proof:**  $\text{FDR}(R^{BH}, P)$

$$= \sum_{j=1}^m (1-\theta_j) \sum_{e=1}^m \frac{1}{e} P(\hat{e} = e, p_j \leq \frac{\alpha e}{m})$$

$$= \sum_{j=1}^m (1-\theta_j) \sum_{e=1}^m \frac{1}{e} \left[ P(\hat{e} \leq e, p_j \leq \frac{\alpha e}{m}) - P(\hat{e} \leq e-1, p_j \leq \frac{\alpha e}{m}) \right] \leq \frac{\alpha e}{m} \text{ by (*)}$$

$$= \sum_{j=1}^m (1-\theta_j) \sum_{e=1}^m \frac{1}{e} \left[ P(\hat{e} \leq e \mid p_j \leq \frac{\alpha e}{m}) - P(\hat{e} \leq e-1 \mid p_j \leq \frac{\alpha e}{m}) \right] \underbrace{P(p_j \leq \frac{\alpha e}{m})}_{\geq P(\hat{e} \leq e-1 \mid p_j \leq \frac{\alpha(p-e)}{m}) \text{ by (\#)}}$$

Hence  $\text{FDR}(R^{\text{BH}}, P)$

$$\leq \frac{\alpha}{m} \sum_{j=1}^m (1-\theta_j) \sum_{\ell=1}^m \left[ P(\hat{\ell} \leq \ell \mid p_j \leq \frac{\alpha \ell}{m}) - P(\hat{\ell} \leq \ell-1 \mid p_j \leq \frac{\alpha(\ell-1)}{m}) \right] \quad \text{telescopic sum!}$$

$$= \frac{\alpha}{m} \sum_{j=1}^m (1-\theta_j) P(\hat{\ell} \leq m \mid p_j \leq \alpha) = \frac{\alpha M_0}{m}$$

How to ensure (#)?

Application 1: positive dependence

[Benjamini and Yekutieli (2001)]



$D_p = \{p \in [0,1]^m : \hat{\ell}(p) \leq \ell-1\}$  is a non decreasing set of  $[0,1]^m$   
i.e.  $\forall p, p' \in [0,1]^m, p \in D_p$  and  $p_j \leq p'_j \forall j$  implies  $p' \in D_{p'}$

(#) holds if

for all  $D \subset [0,1]^m$  non decreasing (measurable)

for all  $j \in \mathcal{B}(P)$

$t \in [0,1] \mapsto P(p \in D \mid p_j \leq t)$  is ↗

weak PRDS

(positively regressively dependent on each one of a subset)

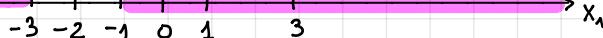
For instance:

\* weak PRADS satisfied in the Gaussian one-sided setting when  $T_{ij} \geq 0$  for all  $ij$

$$\left[ \text{woc for all } j, D_{\{j\}} \mid X_j = y \} = \bigcup_{i=1}^m (\mu_i + T_{ij} (y - \mu_j), \tau_i) \quad \text{indep of } y \right]$$

\* weak PRDS not necessarily satisfied in the Gaussian two-sided setting when  $T_{ij} \geq 0$   
for instance  $m=2, T = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}, \mu = \begin{pmatrix} 0 \\ 2 \end{pmatrix}$  then  $X_2 = X_1 + 2$  a.s.

$$\text{we have } \begin{cases} P(|X_2| > 1 \mid |X_1| \leq 1) = 1 \\ P(|X_2| > 1 \mid |X_1| \leq 3) < 1 \end{cases} \quad \text{at } |X_1 - 2| > 1$$



Open problem: find

$$\sup_{\substack{P \text{ Gaussian} \\ \text{one sided} \\ (\mu, T)}} \{ \text{FDR}(R^{\text{BH}}, P) \} \quad \text{should be not so far from } \alpha$$

show

$$\sup_{\substack{P \text{ Gaussian} \\ \text{two sided} \\ (\mu, T)}} \{ \text{FDR}(R^{\text{BH}}, P) \} \leq \alpha$$

satisfied in cases that are usually 'worst cases'

## Application 2

Martingale dependence

[Storey et al (2004)]

[Heesen Jansen (2015)]

  $D_p = \{ \hat{\ell} \leq p-1 \}$  is measurable with respect to the filtration (with time running backwards)

$$\mathcal{F}_{\hat{\ell}} = \sigma \left( \{1\} p_j \leq \frac{\alpha \ell'}{m} \}, \ell \leq \ell' \leq m, 1 \leq j \leq m \right), \quad 2 \leq \ell \leq m$$

$$\text{Indeed } \hat{\ell} \leq p-1 \Leftrightarrow \forall \ell' > \ell, p(\ell') > \frac{\alpha \ell'}{m} \Leftrightarrow \forall \ell' \geq \ell, \sum_{j=1}^m \mathbb{1}\{p_j \leq \frac{\alpha \ell'}{m}\} < \ell'$$

$$\text{Hence } P(\hat{\ell} \leq p-1 \mid p_j \leq \frac{\alpha(p-1)}{m}) = \mathbb{E} \left[ \mathbb{1}\{\hat{\ell} \leq p-1\} \frac{\mathbb{1}\{p_j \leq \frac{\alpha(p-1)}{m}\}}{P(p_j \leq \frac{\alpha(p-1)}{m})} \right]$$

$$= \mathbb{E} \left[ \mathbb{1}\{\hat{\ell} \leq p-1\} \mathbb{E} \left[ \frac{\mathbb{1}\{p_j \leq \frac{\alpha(p-1)}{m}\}}{P(p_j \leq \frac{\alpha(p-1)}{m})} \mid \mathcal{F}_{\hat{\ell}} \right] \right] \\ = M_{p-1}^{(\hat{\ell})}$$

If  $\boxed{\forall \ell \in \{2, \dots, m\}, \forall j \in \mathcal{B}(P), \mathbb{E}[M_{p-1}^{(\hat{\ell})} \mid \mathcal{F}_{\hat{\ell}}] \leq M_{\hat{\ell}}^{(\hat{\ell})}}$  martingale-type condition

Then  $\leq \mathbb{E}[\mathbb{1}\{\hat{\ell} \leq p-1\} M_{\hat{\ell}}^{(\hat{\ell})}] = P(\hat{\ell} \leq p-1 \mid p_j \leq \frac{\alpha \ell}{m})$  and (#) holds

**Exercise** show that martingale condition is satisfied under independence and directly prove FDR control by using that  $\hat{\ell}$  is a stopping time

A non positive dependence configuration covered by the martingale condition:

Choose  $(p_j, j \in \mathbb{N}_0)$  iid uniform and for  $j \notin \mathbb{B}_0$ , any  $p_j$  in  $[0, \min_{j \in \mathbb{B}_0} p_j]$

Then

$$P(p_j \leq \frac{\alpha(p-1)}{m} \mid \mathcal{F}_{\hat{\ell}}) = P(p_j \leq \frac{\alpha(p-1)}{m} \mid \mathbb{1}\{p_j \leq \frac{\alpha \ell'}{m}\}, \ell \leq \ell' \leq m, 1 \leq j \leq m) \times \mathbb{1}\{p_j \leq \frac{\alpha \ell}{m}\}$$

$$= P(p_j \leq \frac{\alpha(p-1)}{m} \mid p_j \leq \frac{\alpha \ell}{m}, p_{j'} \leq \frac{\alpha \ell'}{m} \text{ pour } j' \neq j \text{ et } j' \in \mathbb{B}_0) \times \mathbb{1}\{p_j \leq \frac{\alpha \ell}{m}\}$$

$$= P(p_j \leq \frac{\alpha(p-1)}{m} \mid p_j \leq \frac{\alpha \ell}{m}) \times \mathbb{1}\{p_j \leq \frac{\alpha \ell}{m}\}$$

car  $p_j \geq p_{j'} \text{ si } j' \notin \mathbb{B}_0$   
by independence inside  $\mathbb{B}_0$  so same or under indep

### ③ Adaptation to dependence

[Lehmann Romano (2005), Romano Wolf (2007), Guo et al (2014)]

- \* FDR control does not really allow to incorporate the dependence (recall  $\text{FDR} = E[\text{FDP}]$ )
- \* we should control the fluctuations of the FDP

New goal: find  $t = \hat{t}$  such that  $\forall P \in \mathcal{P}, P(\text{FDP}(\hat{t}, P) \leq \alpha) \geq 1 - \varsigma$   
often referred to as 'FDP control'

Via a corrected step up procedure: find new critical values  $T_\ell$ ,  $1 \leq \ell \leq m$ ,  
such that for  $\hat{\ell} = \max\{\ell \in \{0, 1, \dots, m\} : p(\ell) \leq T_\ell\}$   
the threshold  $t = T_{\hat{\ell}}$  controls the FDP

Default critical values  $T_\ell = \frac{s(\hat{\ell}) + 1}{m}$ ,  $1 \leq \ell \leq m$  Lehmann Romano critical values

Controls the FDP under weak PRDS assumption [Guo et al (2014)]

**Proposition:** consider p-values satisfying (\*) and  $P \in \mathcal{P}$  satisfying the PRDS assumption  
consider the step-up procedure with Lehmann Romano critical values  
then  $P(\text{FDP}(T_{\hat{\ell}}, P) \leq \alpha) \geq 1 - \varsigma$

**Prof:**  $\{ \text{FDP}(T_{\hat{\ell}}) > \alpha \} = \left\{ \sum_{j=1}^m (1 - \beta_j) \text{ s.t. } p_j \leq T_{\hat{\ell}} \right\} \supseteq \left\{ \sum_{j=1}^m (1 - \beta_j) \text{ s.t. } p_j \leq T_{\hat{\ell}} \text{ and } \hat{\ell} \leq j \leq m \right\} \supseteq \left\{ \exists k \in \{1, \dots, m\} : p_{(k)} \leq \frac{s(\hat{\ell}) + 1}{m} \right\}$   
implies  $m_0 \geq (\hat{\ell}) + 1$

Hence  $P(\text{FDP}(T_{\hat{\ell}}, P) > \alpha) \leq P(\exists k \in \{1, \dots, m_0\} : p_{(k)} \leq \frac{s(\hat{\ell}) + 1}{m_0}) \leq \frac{s(\hat{\ell}) + 1}{m_0}$   
from control of BH( $\varsigma$ ) procedure under the full null □

Better result in Gaussian setting with known  $\Sigma$ ?

Reasoning like [Romano-Shaikh (2006)] we search a bound on  $P(FDP > \alpha)$

$$P(FDP(T_\ell, P) > \alpha) = P\left(\sum_{j=1}^m (1-\vartheta_j) \mathbb{1}\{\hat{P}_j(x) \leq T_\ell\} \geq [\alpha \ell] + 1\right)$$

$$\leq \sum_{\ell=1}^m \mathbb{E}\left[Z_{\ell,\ell} \mid \hat{\ell} = \ell\right]$$

$$\text{where } Z_{\ell,\ell'} = \mathbb{1}\left\{\sum_{j=1}^m \mathbb{1}\{\hat{P}_j(x) \leq T_\ell\} \geq [\alpha \ell'] + 1\right\}$$

$\nearrow$  in  $\ell$   $\searrow$  in  $\ell'$   
whose distribution is known!

$$\leq \sum_{\ell=1}^m \mathbb{E}[Z_{\ell,\ell} \mid \hat{\ell} \geq \ell] - \underbrace{\sum_{\ell=1}^{m-1} \mathbb{E}[Z_{\ell,\ell} \mid \hat{\ell} \geq \ell+1]}_{= \sum_{\ell=2}^m \mathbb{E}[Z_{\ell-1,\ell-1} \mid \hat{\ell} \geq \ell]}$$

$$= \sum_{\ell=1}^m \mathbb{E}\left[\underbrace{(Z_{\ell,\ell} - Z_{\ell-1,\ell-1})}_{\leq (Z_{\ell,\ell} - Z_{\ell-1,\ell-1})_+} \mid \hat{\ell} \geq \ell\right]$$

$$\leq \sum_{\ell=1}^m \mathbb{E}(Z_{\ell,\ell} - Z_{\ell-1,\ell-1})_+ = B((\tau_\ell)_\ell, \alpha, \varsigma, \Gamma)$$

computable  
bound

[Duvallet R. (2015)]

**Proposition:** in the Gaussian setting with known  $\Sigma$ , choose  $T_\ell = \frac{\varsigma}{m} \lceil \alpha \ell \rceil + 1$   
and the step-up procedure with critical values  $\alpha \times T_\ell$ ,  $1 \leq \ell \leq m$   
where  $\alpha$  the largest such that  $B((\tau_\ell)_\ell, \alpha, \varsigma, \Gamma) \leq \varsigma$

Then this procedure controls the FDP

It incorporates dependencies, state of art

**Open problem:** adaptive FDP control with unknown dependence