

## A Direct Approach to False Discovery Rates

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**ERRATA** (as of 4/16/03):

- The following assumptions are made throughout the paper. These are stated in the paper, but unfortunately not very clearly. We assume that  $m$  hypothesis tests are performed based on corresponding p-values  $P_1, P_2, \dots, P_m$ . The realized versions of these p-values are written as  $p_1, p_2, \dots, p_m$ . Let  $H_i = 0$  if null hypothesis  $i$  is true and  $H_i = 1$  if it is false,  $i = 1, 2, \dots, m$ . We assume that  $(P_1, H_1), (P_2, H_2), \dots, (P_m, H_m)$  are i.i.d. random variables such that  $P_i|H_i = 0$  is Uniform(0,1) and  $P_i|H_i = 1$  has continuous probability density function  $g(\cdot)$ . The  $H_i$  are marginally i.i.d. Bernoulli random variables such that  $\Pr(H_i = 0) = \pi_0$  and  $\Pr(H_i = 1) = \pi_1 = 1 - \pi_0$ . It is reasonable to assume that  $P_i|H_i = 1$  is stochastically smaller than  $P_i|H_i = 0$ , although this assumption is not always necessary.

- Page 479, Line -6: The FWER was not necessarily the first multiple hypothesis testing error measure used, but it has been used for many years.

- Page 485, Line +4: This entire paragraph should read:

“Suppose that we enforce the reasonable constraint that  $\hat{\pi}_0 \leq 1$ . The basic point we make here is that using the Benjamini and Hochberg (1995) method to control the FDR at level  $\alpha/\hat{\pi}_0$  is equivalent to (i.e., calls the same  $p$ -values significant as) using the proposed method to control the FDR at level  $\alpha$ . The gain in power from our approach is clear – we control a smaller error rate ( $\alpha \leq \alpha/\hat{\pi}_0$ ), yet reject the same number of null hypotheses.”

- Page 489, Line -6: This sentence should read: “In situations where  $g$  is unknown but Corollary 1 holds with  $g'(1) = 0$ , this estimate is loosely speaking ‘optimal’ in that the bias can usually be made arbitrarily small while obtaining the smallest asymptotic variance (according to standard mle theory) for each  $\lambda$ .”

- Page 492, Section 9: This section in general uses sloppy terminology and (unintentionally) makes exaggerated claims. The section is not about “calculating the optimal  $\lambda$ ” but rather about “estimating the  $\lambda$  that minimizes the mean-squared error.” Therefore, the proposed method calculates an estimate based on a particular criterion that I have labelled as optimal. In later work, others and I have considered other criteria. The important property of Section 9, however, is that I attempt to take into account the variance of a multiple hypothesis testing procedure, which I believe had not been previously considered.