

Report on AOS1910-054

This paper studies the phase transition phenomenon in large scale multiple testing and its implications to large scale genome-wide association studies (GWAS). As a phenomenon, phase transition in large scale multiple testing may no longer be surprising. Nevertheless, the paper provides detailed descriptions of the phenomenon in four different scenarios of multiple independent chi-squared tests and two scenarios of multiple independent two-sided normal test. These results significantly broaden the scope of the phenomenon. As far as I can see, the proofs of the results are all correct. Quite importantly, the paper makes a good connection between these results and large scale GWAS. Basically, the results can be read as saying that a large scale GWAS could be either a complete success or a complete failure, depending on the signal size per sample and the sample size, given the sparsity of significant SNPs. This makes a strong case for careful selection of the sample size, whose importance has been recognized in the field of GWAS for quite a while [2]. I therefore recommend acceptance for publication of the paper after some minor changes.

Comments on the Main Text

1. It would be helpful to emphasize somewhere that in the sub-threshold case, all thresholding procedures fail to get support recovery, whether they control the FWER, FDR, or some other error rate.
2. Section 4.1: It seems that in GWAS on some disorders, a difficulty is that the odds ratio of most disease-associated SNPs is close to one. Therefore, it would be useful to report the asymptotic of $w^2(R)$ as $R \rightarrow 1$.
3. Section 4.2: the relevance of this part needs to be made more clear. The optimal ϕ_1 is for one SNP. Since there can be a large number of disease-associated SNPs, and it may be unknown beforehand which are adversarial, which are protective, and how strong factors they are, how can an optimal ϕ_1 be found / estimated for the entire GWAS?
4. Section 4.2: perhaps it is useful to consider the following. With fixed ϕ_1 , among all SNPs with the same R , which ones are the most significant? This should be easy to answer based on Proposition 4.1, as $w^2(R)$ is a function only in θ_1 .
5. Example 4.1: in Fig. 3, the total number of cases and controls in each panel is half of n . The explanation of the halving is given only later, in Example 4.2. It should be presented in a paragraph on experiment setup before the examples.
6. I would prefer to move Section 6 to the supplement, while moving the current content in the supplement to Section 6.

Comments on the Supplement

1. Proof of Lemma A.1: the proof can and should be simplified by using L'Hôpital's rule,

$$\frac{\int_x^\infty t^{a-1} e^{-t/b} dt}{bx^{a-1} e^{-x/b}} \rightarrow 1, \quad x \rightarrow \infty.$$

2. Eq. (A.4): F^{\leftarrow} has not been defined.
3. Lemma A.2: this is a classical result; see for example, Example 5.3.4 in [3]. No proof is needed.
4. Lemma A.3: the proof can be simplified by using the aforementioned asymptotic based on L'Hôpital's rule.

5. Eq. (B.5): it would be helpful to mention below the display that t_p is the cut-off of an arbitrary thresholding procedure, not just Bonferroni.
6. P. 5: you have all the necessary ingredients, why not just use

$$\begin{aligned} \min_{i \in S} \frac{(Z_\nu(i) + \sqrt{\bar{\Delta}})^2}{u_p} &= [1 + o(1)] \min_{i \in S} \frac{(Z_\nu(i) + \sqrt{\bar{\Delta}})^2}{2 \log p} \\ &\leq [1 + o(1)] \left(\min_{i \in S} \frac{Z_\nu(i)}{\sqrt{2 \log p}} + \sqrt{\bar{r}} \right)^2 \xrightarrow{P} (-\sqrt{1 - \beta} + \sqrt{\bar{r}})^2 < 1 \end{aligned}$$

to directly go from (B.7) to the paragraph starting with “Finally, ...” on p.6. The last limit within the square perhaps should be stated as a lemma as it is also used in the proofs of Theorems 3.4 and 5.2.

7. Eq. (B.19): this is again by L’Hôpital’s rule, not some original result.
8. Eq. (B.23): what does “ $i \in [p]$ ” in the sum stand for?
9. Lemma B.2: this is Theorem 2 in [1] and should be cited as Jaeschke’s theorem; also see [4], p. 600–601.
10. P. 13–14: the proof again can and should be streamlined. I don’t think Lemma B.3 is needed. The lemma itself is a known result (cf. [4], p. 424) combined with the fact that the quantile of the normal distribution is slowly varying at $\pm\infty$.

References

- [1] EICKER, F. (1979). The asymptotic distribution of the suprema of the standardized empirical processes. *Ann. Statist.* **7**, 116–138.
- [2] NISHINO, J., OCHI, H., KOCHI, Y., TSUNODA, T., AND MATSUI, S. (2018). Sample size for successful genome-wide association study of major depressive disorder. *Front. Genet.* **9**:227. doi: 10.3389/fgene.2018.00227.
- [3] RAVISHANKER, N. AND DIPAK, K. D. (2002). *A First Course in Linear Model Theory*. Texts in Statistical Science. Chapman & Hall/CRC.
- [4] SHORACK, G. R. AND WELLNER, J. A. (1986). *Empirical processes with applications to statistics*. John Wiley & Sons, Inc., New York.