

## derivation and explanation of the Bonferroni Correction.Rmd

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This explanation / derivation of the Bonferroni correction follows the Wikipedia article (as of March 16, 2021): [https://en.wikipedia.org/wiki/Bonferroni\\_correction](https://en.wikipedia.org/wiki/Bonferroni_correction) with some commentary added. We are doing  $m$  statistical tests (and for the Bonferroni correction we do not need to assume they are independent). We are going to “declare” that, for any  $i$ , if the **p-value**,  $p_i$  for the  $i^{\text{th}}$  test is  $\leq \tilde{\alpha}$  then that result is **statistically significant**. We want to choose  $\tilde{\alpha}$  such that **the probability of one or more false positives among the  $m$  tests is  $\leq \alpha$**  (a false positive here meaning the null hypothesis was true for test  $i$  but  $p_i \leq \tilde{\alpha}$  so we incorrectly declared this test result to be statistically significant). This is what is meant by **controlling the family-wise error rate (FWER) at level  $\alpha$** . Following the Wikipedia article, we calculate an upper bound for the probability of getting one or more false positives in the following way. By the definition of the p-value, the probability of getting  $p_i \leq \tilde{\alpha}$  just by random chance when the null hypothesis is true is  $\tilde{\alpha}$  (assuming the conditions for validity of the statistical test being used are satisfied). If the null hypothesis is true for only some  $m_0 < m$  of the statistical tests we would only need to control for false positives for that smaller number of tests, but unless we have a good estimate for  $m_0$  we will take the conservative approach of using the number  $m$  of all the tests being done. We will see that the appropriate choice to have the probability that one or more false positives occur be  $\leq \alpha$  is to define

$$\tilde{\alpha} = \alpha/m$$

This is the Bonferroni correction procedure. To verify that this is sufficient, assume the null hypothesis is true for all  $m$  tests. Let  $T_i$  be the event that the p-value for test  $i$  is  $\leq \tilde{\alpha}$  (giving a false positive test result). Then whether or not the tests are independent of each other, the probability that one or more of the  $T_i$  occur is

$$P(\cup_{i=1}^m T_i) \leq \sum_{i=1}^m P(T_i) = m * \tilde{\alpha}$$

The inequality in the equation above is true by the properties of probability measures (see for example [https://en.wikipedia.org/wiki/Boole%27s\\_inequality](https://en.wikipedia.org/wiki/Boole%27s_inequality) ), and the equality in the equation above follows by the definition of a p-value. Hence if we set

$$\tilde{\alpha} = \alpha/m,$$

then the probability of one or more false positives is bounded by  $\alpha$ , so we have controlled the FWER at level  $\alpha$ . So one can either check, for each  $i$ , if  $p_i \leq \alpha/m$  (and declare that test  $i$  is statistically significant if that is true) **OR** equivalently, one can multiply  $p_i$  by  $m$  and declare test  $i$  to be statistically significant if  $m * p_i \leq \alpha$ .

Hope this helps explain how and why Bonferroni correction works.