Let $X_{i1}, \ldots X_{in} \sim N(\mu_i, \sigma^2)$ and $Y_{i1}, \ldots Y_{in} \sim N(\theta_i, \sigma^2)$ independently, for $i = 1, \ldots m$. Think of $X_{i1}, \ldots X_{in}$ as the expression levels of gene *i* for a set of *n* healthy patients, and $Y_{i1}, \ldots Y_{in}$ the expression levels of gene *i* for a set of *n* sick patients. We will test the *m* hypotheses

$$H_i: \mu_i = \theta_i, \quad i = 1, \dots m.$$

Specifically for each gene i, we will compute a t statistic

$$T_{i} = \frac{\bar{X}_{i} - \bar{Y}_{i}}{\sqrt{\frac{1}{n}(S_{X_{i}}^{2} + S_{Y_{i}}^{2})}}$$

,

where $\bar{X}_i = \frac{1}{n} \sum_{j=1}^n X_{ij}$, and $S_{X_i}^2 = \frac{1}{n-1} \sum_{j=1}^n (X_{ij} - \bar{X}_i)^2$, and similarly for $Y_{i1}, \ldots Y_{in}$. We claim that screening out genes from this set in pre-processing, based on total variances, is

We claim that screening out genes from this set in pre-processing, based on total variances, is OK—in that, if we were to compute say FDR estimates using Benjamini-Hochberg on the reduced set, then our estimates would still be valid. For this to be true, we would need to make sure that the distribution of the null t statistics (null p-values) is unchanged after screening. Note that what happens to the distribution of the non-null t statistics (non-null p values) is important for power purposes, but does not concern us when it comes to FDR control.

Hence, it suffices to check that for the null genes i, in which $\mu_i = \theta_i$, the total variance

$$V_i = \frac{1}{2n} \left(\sum_{j=1}^n (X_{ij} - M_i)^2 + \sum_{j=1}^n (Y_{ij} - M_i)^2 \right),$$

where M_i denotes the mean of the joint sample $X_{i1}, \ldots X_{in}, Y_{i1}, \ldots Y_{in}$, is independent of T_i , the t statistic for the *i*th gene. This can be seen in more than one way. From a classical stats perspective: under the null in which the means are the same and therefore $X_{i1}, \ldots X_{in}, Y_{i1}, \ldots Y_{in}$ are drawn independently from the same normal model, M_i is a complete, sufficient statistic for σ^2 . But T_i is ancillary to σ^2 ; this can be seen by dividing numerator and denominator by σ , and noticing that it is then entirely expressed in terms of normal variates within unit variance. Hence by Basu's theorem, the two are independent.