## **Assurance with unknown variance**

Let's consider the same framework as discussed so far, but with unknown variance  $\sigma^2$  that is equal for both groups ( $\sigma = \sigma_0 = \sigma_1$ )

In this case we typically base our test on the t-distribution rather than the normal distribution, i.e. we reject the null hypothesis if

$$\overline{y}_1 - \overline{y}_0 > t_{1-\alpha,2n-2} \hat{\sigma}_{\sqrt{\frac{2}{n}}},$$

with  $\hat{\sigma}^2 = \frac{\sum_{i=0}^{1} \sum_{j=1}^{n_i} (y_{ij} - \bar{y}_i)^2}{2n-2}$  the estimate of the variance and  $t_{1-\alpha,2n-2}$  the 1- $\alpha$  percentile of the *t*-distribution with 2n-2 degrees of freedom.

### Since we do now not know $\sigma$ , we may also wish to define a prior on $\sigma$ .

O'Hagan, A., Stevens, J. W., & Campbell, M. J. (2005). Assurance in clinical trial design. *Pharmaceutical Statistics*, 4(3), 187–201.

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# For the case of unknown sigma (or complex success criteria) a simulation approach can be used to calculate assurance

- 1. Set counters I = P = 0. Set required number of simulations N.
- 2. Sample  $\delta$  and  $\sigma$  from their joint prior distribution.
- 3. Sample an observed effect  $\bar{y}_1 \bar{y}_0 \sim N(\delta, \tau)$  and an estimated standard deviation using  $(2n 2)\hat{\sigma}^2/\sigma^2 \sim \chi^2_{2n-2}$ .
- 4. Increment *P* if  $\overline{y}_1 \overline{y}_0 > t_{1-\alpha,2n-2} \hat{\sigma} \sqrt{\frac{2}{n}}$ .
- 5. Increment *I*. If I < N, go to step 2.
- 6. Estimate assurance by *P/N*



## A comment on priors on the standard deviation

Both in my own experience and that of Walley et al (2015) a prior on  $\sigma$  does not affect assurance in a relevant manner compared to calculating assurance based on the mean of the prior for  $\sigma$ 

This finding is due to the assurance value changing almost linearly over the credible range for  $\sigma$ , which when combined with an approximately symmetrical prior distribution results in the marginal assurance value almost being equal to the assurance at the prior mean for  $\sigma$ .





Walley, R. J., Smith, C. L., Gale, J. D., & Woodward, P. (2015). Advantages of a wholly Bayesian approach to assessing efficacy in early drug development: A case study. *Pharmaceutical Statistics*, *14*(3), 205–215. https://doi.org/10.1002/pst.1675

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