

Deriving the Power of a Test

Consider a situation in which we would like to test whether an overdose rate has shrunk from p_{old} to $p_{\text{new}} = p_{\text{old}}/2$ (ie: whether the overdose rate has halved). The question is – how large does our sample need to be to get a certain power for a test of a certain size?

Strictly speaking, we should fit a binomial generalized linear model and derive a test statistics from that model. Given the large number of sample, however, we take a “back of the envelope” approach instead.

Let the original overdose rate be p_{old} , the new rate p_{new} , and let n individuals be sampled in each case (in this case, 15, 000) or which d_{old} and d_{new} die of overdose. The hypotheses we want to test are

$$\begin{aligned} H_0 : p_{\text{new}} &= p_{\text{old}} \\ H_1 : p_{\text{new}} &= p_{\text{old}} / 2 \end{aligned}$$

Now, note that

$$d_{\text{old}} \sim \text{Bin}(n, p_{\text{old}}) \Rightarrow \text{Var}(d_{\text{old}}) = np_{\text{old}}(1 - p_{\text{old}}) = np_{\text{old}}(1 - p_{\text{old}})$$

Since the rates are very small and n is very large, we approximate

$$d_{\text{old}} \sim N(np_{\text{old}}, np_{\text{old}}\bar{p}_{\text{old}})$$

and similarly for d_{new} . This means that

$$d_{\text{old}} - d_{\text{new}} \sim N(n\{p_{\text{old}} - p_{\text{new}}\}, n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}}))$$

Our test statistic will therefore be

$$Z = \frac{d_{\text{old}} - d_{\text{new}}}{\sqrt{n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}} \sim \begin{cases} N(0,1) & \text{under } H_0 \\ N(np_{\text{new}}, 1) & \text{under } H_1 \end{cases}$$

Now, if we want the significance of our test to be α ,

$$\begin{aligned} \mathbb{P}(H_0 \text{ rejected} \mid H_0 \text{ true}) &= \alpha \\ \mathbb{P}\left(\frac{d_{\text{old}} - d_{\text{new}}}{\sqrt{n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}} > C\right) &= \alpha \\ C &= z_\alpha \end{aligned}$$

All that said, the power of our test, $1 - \beta$, is given by

$$\begin{aligned} \mathbb{P}(H_0 \text{ accepted} \mid H_1 \text{ true}) &= \beta \\ \mathbb{P}\left(\frac{d_{\text{old}} - d_{\text{new}}}{\sqrt{n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}} > C\right) &= 1 - \beta \\ \mathbb{P}\left(\frac{d_{\text{old}} - d_{\text{new}} - np_{\text{new}}}{\sqrt{n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}} > C - \frac{np_{\text{new}}}{\sqrt{n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}}\right) &= 1 - \beta \\ C - \frac{np_{\text{new}}}{\sqrt{n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}} &= z_{1-\beta} \end{aligned}$$

Solving to eliminate C

$$\begin{aligned} z_{\alpha} - z_{1-\beta} &= \frac{np_{\text{new}}}{\sqrt{n(p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}} \\ \frac{(z_{\alpha} - z_{1-\beta})^2 (p_{\text{old}}\bar{p}_{\text{old}} + p_{\text{new}}\bar{p}_{\text{new}})}{p_{\text{new}}^2} &= n \end{aligned}$$