DRAFT 0.1

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Note on Finding a Measurement from a Limit

Suppose we are given an upper limit s_{up} for the expected number of signal events s at confidence level $1 - \alpha$, as well as the expected limit assuming s = 0, $E[s_{up}|0]$ and "error band", i.e., the standard deviation $\sigma[s_{up}|0]$. The goal is to find from these values an estimate \hat{s} for s as well as its standard deviation $\sigma_{\hat{s}}$. In contrast to the method of Marumi, the result below only uses the observed and expected limit, not the width of the error band.

The reported limit was based on the p-value

$$p_s = P(\hat{s} \le \hat{s}_{\text{obs}} | s) . \tag{1}$$

If \hat{s} is Gaussian distributed with mean s and variance $\sigma_{\hat{s}}^2 = V[\hat{s}|s]$, then this is

$$p_s = \Phi\left(\frac{\hat{s}-s}{\sigma_{\hat{s}}}\right) \,, \tag{2}$$

where Φ is the standard Gaussian cumulative distribution. Here \hat{s} refers to the observed value; we drop "obs" to simplify the notation. The upper limit is the value of s such that $p_s = \alpha$, which gives

$$s_{\rm up} = \hat{s} + \Phi^{-1} (1 - \alpha) \sigma_{\hat{s}|s_{\rm up}} , \qquad (3)$$

where $\sigma_{\hat{s}|s_{up}}^2 = V[\hat{s}|s_{up}]$ is the variance of \hat{s} under assumption of $s = s_{up}$ and Φ^{-1} is the standard Gaussian quantile.

Solving Eq. (3) for \hat{s} gives

$$\hat{s} = s_{\rm up} + Q\sigma_{\hat{s}|s_{\rm up}} , \qquad (4)$$

where we define $Q = \Phi^{-1}(1-\alpha)$ (e.g., Q = 1.64 for a confidence level $1-\alpha = 0.95$). This does not quite give the desired answer because the standard deviation $\sigma_{\hat{s}|s_{up}}$ is not given. Rather, we have the expected s_{up} as well its the standard deviation of s_{up} , under the assumption of s = 0. We will denote these below as $E_0 \equiv E[s_{up}|0]$ and $\sigma_0 = \sigma[s_{up}|0]$.

To relate the given ingredients — s_{up} , E_0 and σ_0 — to the desired results, suppose the estimate of s results from a counting experiment where a number of events n follows a Poisson distribution with mean s + b. Here the expected number of background events b was known when the limit was initially found but it is not an input for the present problem. The maximum-likelihood estimator for s and its variance are

$$\hat{s} = n - b , \qquad (5)$$

$$V[\hat{s}|s] = s+b. (6)$$

Using Eq. (6) for the variance allows us to rewrite Eq. (3) so as to relate s_{up} and \hat{s} ,

$$(s_{\rm up} - \hat{s})^2 = Q^2(s_{\rm up} + b) .$$
(7)

Solving this equation for $s_{\rm up}$

$$s_{\rm up} = \hat{s} + \frac{Q^2}{2} \left[1 + \left(1 + \frac{4(\hat{s} + b)}{Q^2} \right)^{1/2} \right] \,. \tag{8}$$

Or solving instead for \hat{s} gives

$$\hat{s} = s_{\rm up} - Q\sqrt{s_{\rm up} + b} \ . \tag{9}$$

We are given the expected limit assuming s = 0, here written E_0 . Using Eq. (8) and approximating $E[\sqrt{\hat{s}}] \approx \sqrt{s}$ and then evaluating with s = 0 gives

$$E_0 \equiv E[s_{\rm up}|s=0] \approx \frac{Q^2}{2} \left[1 + \left(1 + \frac{4b}{Q^2}\right)^{1/2} \right] \,. \tag{10}$$

The other given input is σ_0 , the standard deviation of s_{up} assuming s = 0. By using error propagation with Eq. (8) we can then find the variance of s_{up} in terms of the variance of \hat{s} , and then evaluate at s = 0. This gives

$$\sigma_0 \equiv \sigma[s_{\rm up}|s=0] = \sqrt{b} \left[1 + \left(1 + \frac{4b}{Q^2}\right)^{-1/2} \right] \,. \tag{11}$$

Both Eqs. (10) and (11) can be solved in principle for b, and so the system is over-determined. Equation (10) for the expected limit can easily be solved in closed form; the equation for the standard deviation (11) is quartic and thus less easy to handle. If we ignore the information from σ_0 and just solve Eq. (10) for b we obtain

$$b = \frac{E_0^2}{Q^2} - E_0 . (12)$$

So by first solving for b and using it in Eq. (9) one can find the estimator \hat{s} . Using these together with Eq. (6) gives its standard deviation

$$\sigma_{\hat{s}} = \sqrt{\hat{s} + b} \ . \tag{13}$$