

Error Analysis

Good statistics website

<http://faculty.vassar.edu/lowry/webtext.html>

see Salter paper for LS analysis

Propagation of errors

rules for error propagation pg56 Table 3.1 in Harris

addition or subtraction

$$a + b + c = d \quad a - b - c = d$$

$$s_d = (s_a^2 + s_b^2 + s_c^2)^{1/2}$$

multiplication

$$a*b/c = d$$

$$s_d/d = [(s_a/a)^2 + (s_b/b)^2 + (s_c/c)^2]^{1/2}$$

exponents

$$y = x^a$$

$$s_y/y = a s_x/x$$

logs

$$y = \ln x$$

$$s_y = s_x/x$$

$$y = e^x$$

$$s_y/y = s_x$$

Least squares analysis

The “best fit” line

Minimize the square of the vertical deviation - d_i^2
$$d_i^2 = (y_i - y)^2 = (y_i - (mx_i + b))^2$$

After some calculus....

$$\text{Slope} = [n\Sigma(x_i y_i) - \Sigma x_i \Sigma y_i] / D$$

$$\text{Where } D = n\Sigma(x_i^2) - (\Sigma x_i)^2$$

$$\text{Intercept} = [\Sigma(x_i^2)\Sigma y_i - \Sigma x_i \Sigma(x_i y_i)] / D = y_{\text{ave}} - mx_{\text{ave}}$$

Error analysis using Least Squares

Standard Error of Regression or standard error of residuals

The sum of the squares of the vertical deviates

$$\begin{aligned} s_y &= [\Sigma(d_i - d_{\text{av}})^2 / (N-2)]^{1/2} = [\Sigma(d_i)^2 / (N-2)]^{1/2} \\ &= [\Sigma(y_i - mx_i - b)^2 / (N-2)]^{1/2} \end{aligned}$$

$$\text{error in slope} = s_m = [ns_y^2 / D]^{1/2}$$

$$\text{error in intercept} = s_b = [s_y^2 \Sigma(x_i^2) / D]^{1/2}$$

For typical calibration plot:

error in unknown x from interpolation

$$s_x = \left\{ (s_y/m)^2 \left[1/k_{\text{unk}} + 1/D (nx_{\text{int}}^2 - 2x_{\text{int}}\Sigma x_i + \Sigma(x_i^2)) \right] \right\}^{1/2}$$

where k = number of replicate measurements of the unknown

For standard addition plot:

error in x intercept

$$s_x = \left\{ (s_y/m)^2 \left[1/D (nx_{\text{int}}^2 - 2x_{\text{int}}\Sigma x_i + \Sigma(x_i^2)) \right] \right\}^{1/2}$$

