

Errors

When you make a single measurement, you get a value and an uncertainty:  $x \pm \delta x$

$\delta x$  should be your best estimate of the standard deviation of the probability distribution for  $x$  (see below)

If you measure  $x$  many times:  $x_1, x_2, x_3, \dots$

then the best answer for  $x$  is  $\bar{x} = \sum_{i=1}^N x_i$

But what is the uncertainty?

The standard deviation:

$$\sigma = \sqrt{\frac{\sum_{i=1}^N (\bar{x} - x_i)^2}{N-1}}$$

should match  $\delta x$  above. Because we've made more measurements now, the uncertainty of the true value in  $x$  should be smaller.

It is:  $\sigma_m = \frac{\sigma}{\sqrt{N}}$

So repeating measurements gives a more accurate determination of  $x$  and also information about the probability distribution  $P_x(x)$ .

It's easy to see that  $\sigma_m$ , the standard error of the mean should be  $\frac{\sigma}{\sqrt{N}}$ .

Imagine  $N=2$ .

$$\bar{x} = (x_1 + x_2) / 2$$

and the uncertainty in each of  $x_1$  and  $x_2$  is  $\delta x$ .

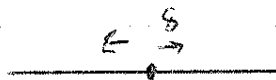
So the uncertainty in  $\bar{x}$  is  $\frac{\sqrt{(\delta x)^2 + (\delta x)^2}}{2} = \frac{\delta x}{\sqrt{2}}$

The same argument shows it is  $\frac{\delta x}{\sqrt{N}}$  for  $N$  measurements.

## Quadrature

Why do we add errors in quadrature?

Imagine a random walk: Consider an object starting at the origin that takes steps either left or right, of size  $S$ .



Label each step:

$$\Delta x_1 = \pm S$$

$$\Delta x_2 = \pm S$$

$\vdots$

$x_i$  = position after  $i$  steps

$$x_N = \sum_{i=1}^N \Delta x_i$$

After  $N$  steps, where do we expect to be?

$$\langle x_N \rangle = \sum_{i=1}^N \langle \Delta x_i \rangle$$

$$= 0$$

if  $+s$  and  $-s$  are  
equally likely.

But on any given run, we are unlikely to end up exactly at 0.

How far from the starting point do we expect to be?

$$\text{Consider: } \langle x_N^2 \rangle = \left\langle \left( \sum_{i=1}^N \Delta x_i \right)^2 \right\rangle$$

$$\langle x_1^2 \rangle = \langle \Delta x_1^2 \rangle = s^2$$

$$\begin{aligned} \langle x_2^2 \rangle &= \langle (\Delta x_1 + \Delta x_2)^2 \rangle \\ &= \langle (\Delta x_1^2 + 2\Delta x_1 \Delta x_2 + \Delta x_2^2) \rangle \end{aligned}$$

if the steps are independent (ie  $\Delta x_2$  doesn't depend on  $\Delta x_1$ ), then  $\langle \Delta x_1 \Delta x_2 \rangle = 0$

$$\text{so } \langle x_2^2 \rangle = 2s^2$$

$$\text{and } \langle x_N^2 \rangle = Ns^2$$

$$\text{so that } \sqrt{\langle x_N^2 \rangle} = \sqrt{N} s$$

$$\text{and } \frac{\sqrt{\langle x_N^2 \rangle}}{N} = \frac{s}{\sqrt{N}}$$

Errors are similar. You can think of each step as

corresponding to the error in a measurement. If we do  $N$  measurements, the error in the sum does increase  $\sim \sqrt{N} S$ , but when the average is taken, the error decreases:  $S/\sqrt{N}$ .

## Quadrature part II

Imagine measuring quantities  $x$  and  $y$  where we want  $z = x + y$

Assume the errors in  $x$  and  $y$  are Gaussian distributed and independent.

$$\text{So } P_x(x) = \frac{1}{\sqrt{2\pi} \sigma_x} e^{-\frac{(x-x_0)^2}{2\sigma_x^2}}$$

$$\text{and } P_y(y) = \frac{1}{\sqrt{2\pi} \sigma_y} e^{-\frac{(y-y_0)^2}{2\sigma_y^2}}$$

$x_0$  and  $y_0$  are the true values and  $\sigma_x$  and  $\sigma_y$  are the widths of the distributions.

Make many measurements of  $x$  and  $y$ .

$$\text{now: } z_i = x_i + y_i$$

What does  $P_z(z)$  look like?

$$\text{Claim: } P_z(z) = \frac{1}{\sqrt{2\pi} \sigma_z} e^{-\frac{(z-z_0)^2}{2\sigma_z^2}}$$

$$\text{with } \sigma_z = \sqrt{\sigma_x^2 + \sigma_y^2}$$

$$z_0 = x_0 + y_0$$

Proof:

Since the errors are independent:

$$P_z(z) = \int_{-\infty}^{\infty} P_x(x) P_y(z-x) dx$$

There are many ways to arrive at any particular value of  $z$ . We integrate over all possible  $x, y$  pairs that give  $z = x + y$ .

$$P_z(z) = \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_x\sigma_y} e^{-\frac{(x-x_0)^2}{2\sigma_x^2}} e^{-\frac{(z-x)^2}{2\sigma_y^2}}$$

Strategy: expand the arguments of the exponentials, then complete the square.

So we have something like:

$$\frac{1}{2\pi\sigma_x\sigma_y} \int_{-\infty}^{\infty} e^{-\frac{(x-A)^2}{2B^2}} f(z) dx$$

$$\text{then: } \int_{-\infty}^{\infty} e^{-\frac{(x-A)^2}{2B^2}} = \sqrt{2\pi} B$$

after a bit of algebra you can prove the claim above.

This is exactly what was demonstrated in homework 3.