

Recitation 1: Uncertainty and Error Propagation

8.S50

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Basic Statistics

Average

$$\langle x \rangle = \frac{1}{N} \sum_{i=1}^N x_i$$

Variance

$$s_x^2 = \frac{1}{N-1} \sum_{i=1}^N (x_i - \langle x \rangle)^2$$

“Standard Deviation”

s_x or σ

Uncertainty

- ▶ Any representation of these three measurements are meaningless without uncertainty
- ▶ A physical measurement has two components: the best value and its uncertainty

Uncertainty or *error* is the estimate of how close our answer is to the true value.

Types of Uncertainty

Systematic	Statistical
Systematic Imperfection in <ul style="list-style-type: none">• Method• Apparatus• Calibration• Environment• Model/Hypothesis	Random/Statistical Variation <ul style="list-style-type: none">• Underlying physical process<ul style="list-style-type: none">◦ e.g. Poisson fluctuations• Instrument noise
Not independent from measurement to measurement	Independent from measurement to measurement <ul style="list-style-type: none">• “averages out” when repeating the measurement
Limits accuracy	Limits Precision

Measurements of Uncertainty

- ▶ Standard deviation σ is often used as the estimate of uncertainty
- ▶ Standard Deviation of the Mean (Standard Error) $SE = \frac{\sigma}{\sqrt{N}}$ can also be reported for the uncertainty in the mean.
- ▶ For counting experiments with N observed events, the estimated uncertainty on the mean of the underlying Poisson distribution is \sqrt{N}

Error Propagation (Propagation of Uncertainty)

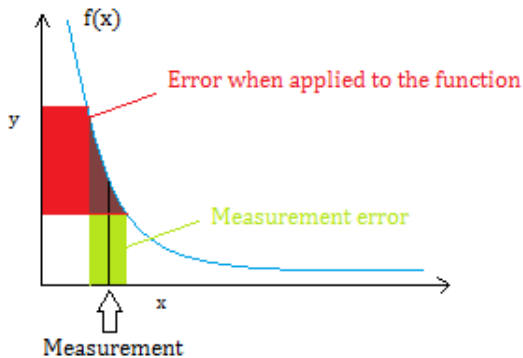
- ▶ Often, our experiment will measure one observable, but we will want to report a quantity derived from the observable.
- ▶ Also, often the reported final result will depend on multiple observables.
- ▶ Finally, the uncertainty on some observables may have contributions from multiple sources.
- ▶ We therefore need to understand how to propagate uncertainties on various components to the final result.

Error Propagation

- ▶ Suppose we measure two observables. $u \pm \sigma_u$ and $v \pm \sigma_v$
- ▶ Say we want to report the uncertainty of a measurement that is dependent on both $x = f(u, v)$
- ▶ Think of a Taylor expansion. (with σ_u, σ_v small and independent):

$$\sigma_x^2 = \sigma_u^2 \left(\frac{\partial x}{\partial u} \right)^2 + \sigma_v^2 \left(\frac{\partial x}{\partial v} \right)^2$$

Error Propagation



Error Propagation

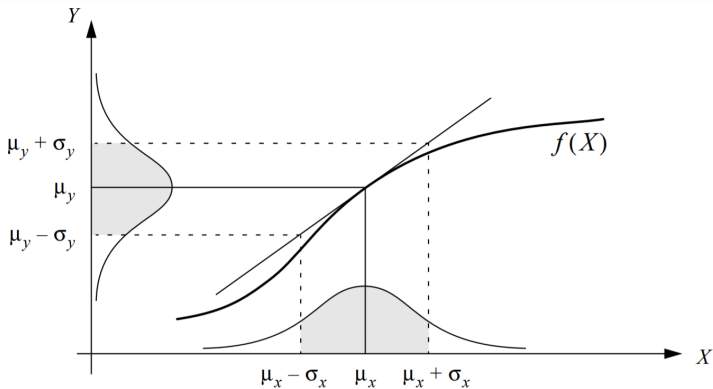


Figure 2: One-dimensional case of a nonlinear error propagation problem

Error Propagation

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Propagation Rules - Sum and Difference

- Absolute uncertainties *add*

$$\underline{x = u \pm v}$$

$$\rightarrow \frac{\partial x}{\partial u} = \frac{\partial x}{\partial v} = 1$$

$$\sigma_x^2 = \sigma_u^2 + \sigma_v^2$$

$$\boxed{\sigma_x = \sqrt{\sigma_u^2 + \sigma_v^2}}$$

Propagation Rules - Products or Quotients

- Fractional uncertainties *add*

$$\underline{x = auv}$$

$$\rightarrow \frac{\partial x}{\partial u} = av, \frac{\partial x}{\partial v} = au$$

$$\frac{\sigma_x^2}{x^2} = \frac{\sigma_u^2}{u^2} + \frac{\sigma_v^2}{v^2}$$

$$\sigma_x = x \sqrt{\left(\frac{\sigma_u}{u}\right)^2 + \left(\frac{\sigma_v}{v}\right)^2}$$

Propagation Rules - Powers

- Fractional uncertainties *multiply* by power

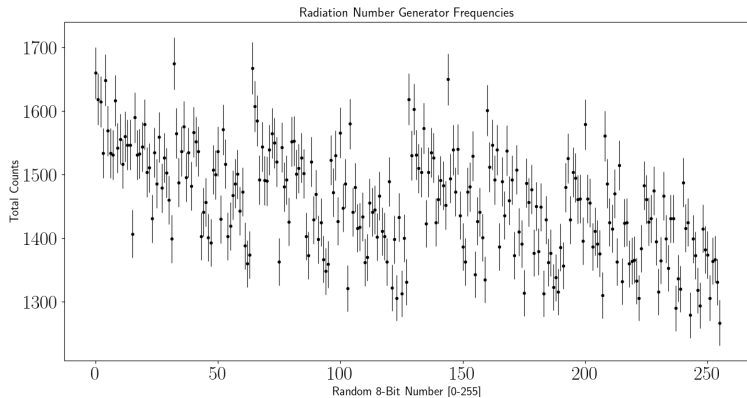
$$\underline{x = au^p}$$

$$\rightarrow \frac{\partial x}{\partial u} = apu^{p-1} = p \frac{x}{u}$$

$$\frac{\sigma_x}{x} = p \frac{\sigma_u}{u}$$

$$\boxed{\sigma_x = xp \frac{\sigma_u}{u}}$$

Representing Uncertainty



Representing Uncertainty

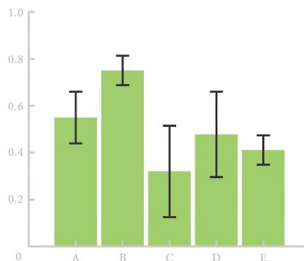
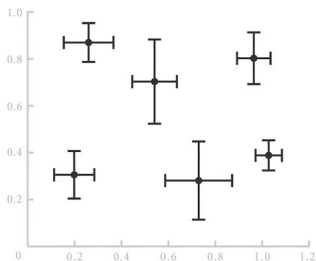
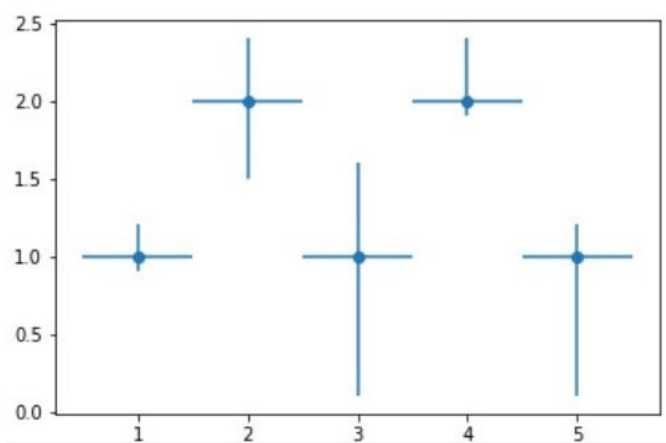


Figure: The Data Visualization Catalog

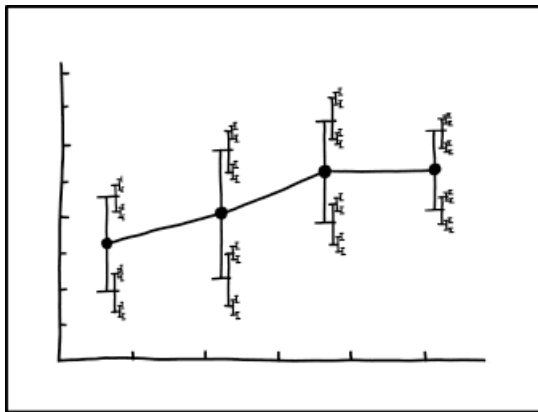
matplotlib.pyplot.errorbar

```
1 import matplotlib.pyplot as plt
2
3 x =[1, 2, 3, 4, 5]
4 y =[1, 2, 1, 2, 1]
5
6 y_errormin =[0.1, 0.5, 0.9,
7              0.1, 0.9]
8 y_errormax =[0.2, 0.4, 0.6,
9              0.4, 0.2]
10
11 x_error = 0.5
12 y_error =[y_errormin, y_errormax]
13
14 # plotting graph
15 # plt.plot(x, y)
16 plt.errorbar(x, y,
17              yerr = y_error,
18              xerr = x_error,
19              fmt = 'o')
```


pyplot errorbars



Relevant XKCD



I DON'T KNOW HOW TO PROPAGATE
ERROR CORRECTLY, SO I JUST PUT
ERROR BARS ON ALL MY ERROR BARS.