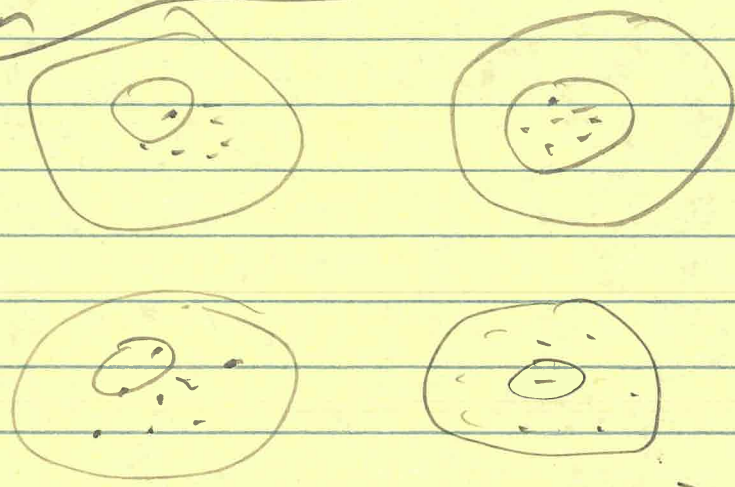


- 1) more about error
- 2) python

321
 cut top 15
 14.6
 15.4
 L3.2

(precision)



(accuracy)

Info for flipped Taylor Chap 4

(note students to suggest labels for plot above)

→ "accuracy" involves knowing what is a good value ←

1) Can reduce statistical error (inverse precision) by taking more measurements NT. Note example for LHC Higgs

⇒ Born experiment [last up] actually

2) must address accuracy, by removing and/or reducing "systematic error"

⇒ give example

took data

→ 125.3 ± 0.6 CMS
 and 126.0 ± 0.6 ATLAS
 (16eV/c)

391
L2.1b

(Example of reducing σ)

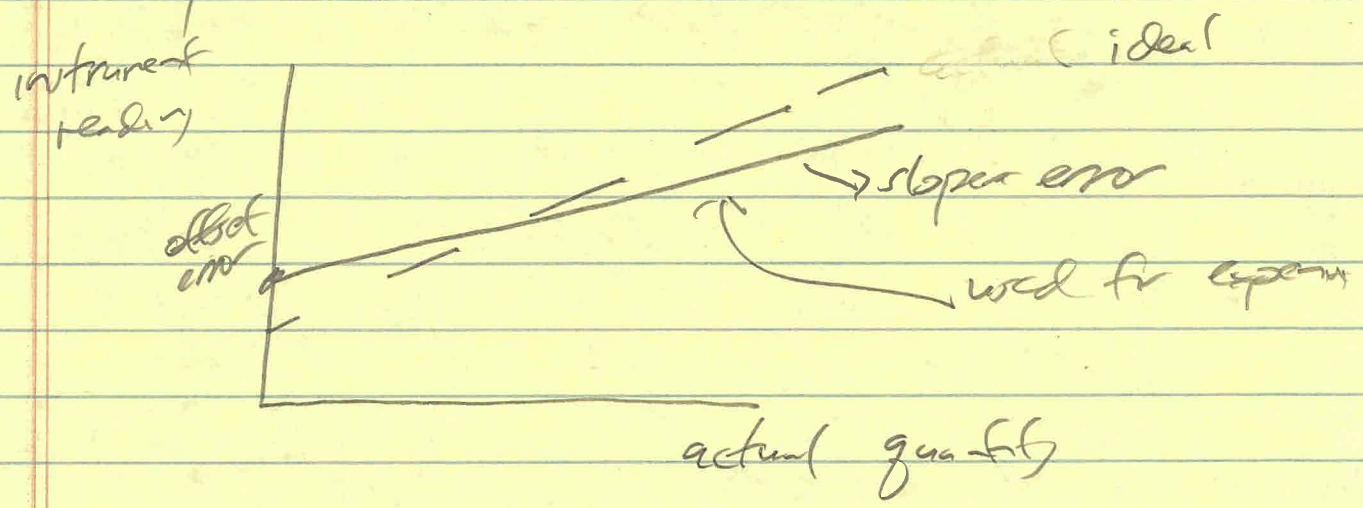
Higgs Boson, took ⁴⁰/₄ year
took ^{enough} data to where probability of getting
at least as strong a result
(for alt expl) was 1:3 million
and significance of 5 sigma (σ)

results were * (125.3 ± 0.6) GeV/c \rightarrow CMS
 (126.4 ± 0.6) GeV/c \rightarrow ATLAS

Teams were blinded from each other since 2011

Example of systematic error:

Example would be mis-calibrated instrument



- 1) Normal distributions - PDFs + the Gaussian
- 2) SDOM - standard deviation of the mean
- ~~3) weighted averages~~

3) Lab 2 Brownian Motion

- 1) typically ^{plan for} ~~with~~ N measurements of same thing for an experiment.
 (then we might change an variable and go again, that's later)

$$x_1, x_2, x_3, \dots, x_N$$

→ our goal, then, is to develop a "best estimate" & "best width estimate" from these N

We start our statistical analysis process by assuming our distribution of means x_i has a well-defined mean + standard deviation \bar{x} + σ .

Thus the central limit theorem says that M sets of these N means would be "normal", →

Look at subonline + 2 results

39
~~40~~
~~41~~
~~42~~
 3.7

first will look at 1) increasing precision

a) consider distribution of ¹⁰ measurements

26, 29, 26, 28, 27, 29, 25, 28, 26, 25

want to look at the distribution, so we

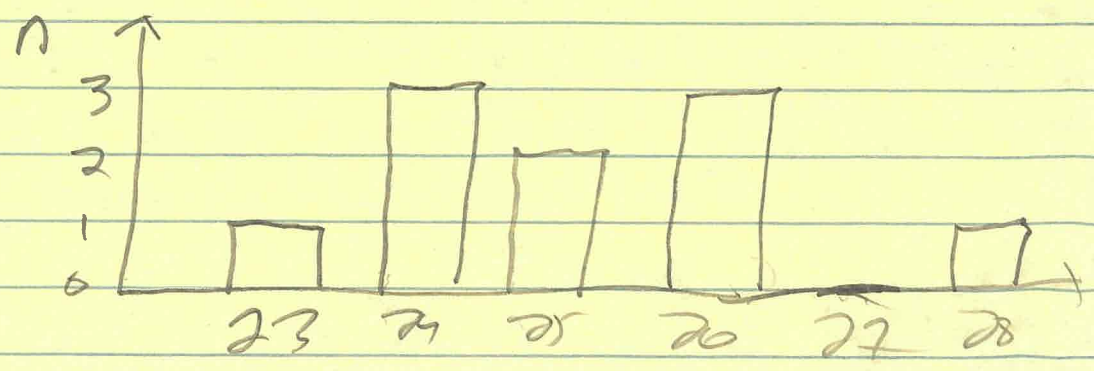
i) sort the numbers

23, 24, 24, 24, 25, 25, 26, 26, 27, 28

ii) count

23	24	25	26	27	28
1	3	2	3	1	1

iii) plot frequency of occurrence



3.11
 $F(x)$
~~3.3~~
 3.3

iv) represent as mean

$$\bar{x} = \frac{\sum x_i}{N} = \frac{\sum_{j=1}^m n_j x_j}{\sum n_j} \quad (\Rightarrow N)$$

- j ranges over bins
- x_j value of b_j
- n_j # in bin

v) could write $f_j = \frac{n_j}{N}$

$$\Rightarrow \bar{x} = \sum_j f_j x_j$$

b) aside about real world "bin widths"

normally measure real number with some accuracy, say 0.1 and

distribution of 50 measure looks instead like the [already ordered]

23.7, 23.8, 23.9, 23.9, 27.6, etc

~~~~~  
 0.1    0.1    0.0    0.7

look at spaces in data

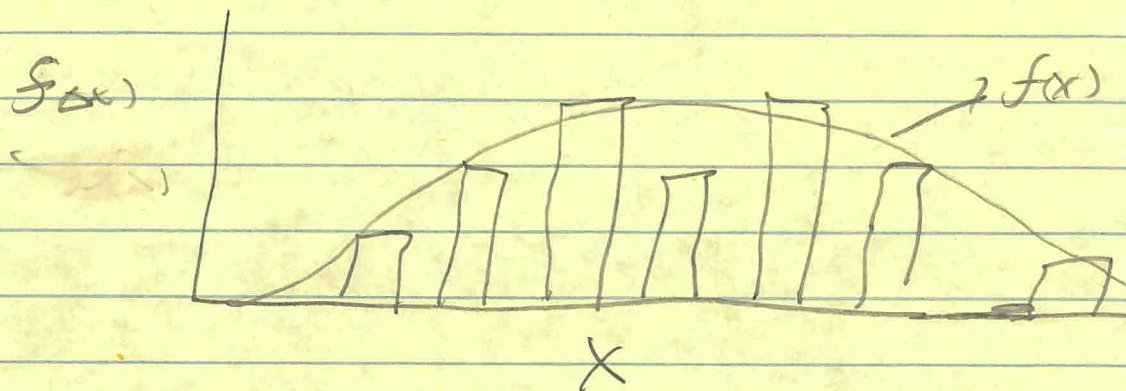
pick some reasonable average of gaps or "bin widths"

3π  
~~6π~~  
 3π  
 2π

(i) were working towards a statistical description of data distribution

$f(x)$  a probability distribution-fcn (it is data dependent)

so will come from



limit as  $\Delta x \rightarrow 0, N \rightarrow \infty$  is  $f(x)$

→ then  $f(x) dx$  is the fraction of samples between  $x$  and  $x+dx$

or more properly formally

$$\int_a^b f(x) dx = \text{fraction of samples between } a \text{ + } b$$

d) need to normalize s.t.

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

3.1  
~~with FE~~  
~~1.25~~  
 3.5

e) now, with our  $f(x)$  properly normalized we can state that:

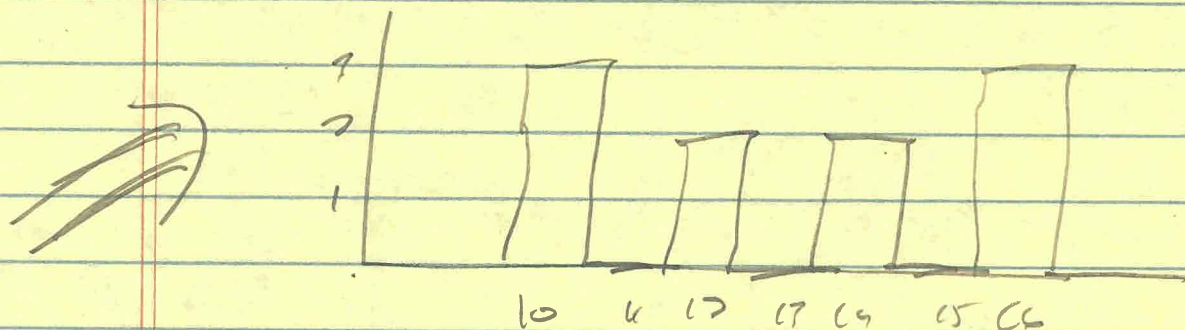
these two things define our dist. distribution somewhat well

i)  $\bar{x} = \int_{-\infty}^{\infty} x f(x) dx$  "expectation value"

ii)  $\sigma^2 = \int_{-\infty}^{\infty} (x - \bar{x})^2 f(x) dx$  "variance"

f) Central Limit Theorem

start with any distribution with well-defined mean and variance



can set a sample size  $N$

and take  $N$  samples (example) of above

set  $N=4 \Rightarrow [10, 10, 14, 16] \Rightarrow \bar{x}_1 = 12.5$

keep doing this, then plot freq. dist of example means

391  
~~391~~  
~~391~~  
~~391~~  
391

get something that looks like a "normal distribution"

further, if we increase sample size  $N$  and look at freq. distr. looks more normal

CLT says that as  $N \rightarrow \infty$  freq. distr.  $\rightarrow$  "normal" with well-defined mean & variance.

note, we started with any distribution and only assumed it had a well defined mean and variance

It might not, itself, be "normal"

g) but what does it mean to be "normal"?

implies that  $f(x) \approx$  Gaussian  $= e^{-\frac{x^2}{2\sigma^2}}$   
or  $e^{-\frac{(x-\bar{x})^2}{2\sigma^2}}$



311  
~~1000~~  
~~1000~~  
~~1000~~  
 L273.7

for now we'll call  $b$  the "width func"  
 and  $\bar{x}$  the "maximally probable value"

→ would typically need to normalize this  
 $f(x)$  s.t.

$$\int_{-\infty}^{\infty} f(x) dx = 1$$

h.) ERF

so the probability of finding a sample  
 between  $\mu - b$  &  $\mu + b$  is

$$\text{Prob}(\mu - b, \mu + b) = \int_{\mu - b}^{\mu + b} \frac{1}{\sqrt{2\pi} b} e^{-\frac{(x - \mu)^2}{2b^2}} dx$$

can let  $z \equiv \frac{x - \mu}{b} \Rightarrow$

$$\text{Prob} = \int_{-1}^1 \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz$$

called the "error func" or ERF

look up when in power

## 2) SDom

have established, given certain (reasonable) assumptions,  $\bar{X} = \bar{x}$  and  $b_b = b$

still working w/  $N$   $x_i$ 's, what is "reliability" of  $\bar{X} = \bar{x}$ ?

to consider, look at  $\bar{x}_k = \frac{\sum_{i=1}^N x_{i,k}}{N}$   $k=1, \dots, M$   
 dist. of  $\bar{x}_k$  over  $M$  "sets" (sets)

we note that if  $x_1, x_2, \dots, x_N$  is normally distributed then  $\bar{x}$  is normally distributed (simple fcn of  $x_i$ 's)

2<sup>nd</sup>, each  $x_i$  has the same true value  $\bar{X}$

$$\text{so } \bar{X} = \frac{\bar{x}_1 + \bar{x}_2 + \bar{x}_3 + \dots}{N} = \text{true value}$$

Now look at stability 2 results

So measuring  $M$  sets doesn't change  $\bar{X} = \bar{x}$  (of course)

(D): what is width estimate of  $\bar{X}$ ? (add errors in  $\bar{x}$  in quadrature)

$$\sigma_{\bar{X}} = \sqrt{\left(\frac{\partial \bar{X}}{\partial x_1} \sigma_{x_1}\right)^2 + \left(\frac{\partial \bar{X}}{\partial x_2} \sigma_{x_2}\right)^2 + \dots + \left(\frac{\partial \bar{X}}{\partial x_N} \sigma_{x_N}\right)^2}$$

but

$$\sigma_{x_1} = \sigma_{x_2} = \dots = \sigma_{x_N} = \sigma_x$$

391  
F(6 ~~5~~)  
L3.5

and:  $\frac{\partial \bar{x}}{\partial x_1} = \frac{\partial \bar{x}}{\partial x_2} = \dots = \frac{\partial \bar{x}}{\partial x_N} = \frac{1}{N}$

$$\Rightarrow \sigma_{\bar{x}} = \sqrt{\left(\frac{1}{N} \sigma_x\right)^2 + \dots + \left(\frac{1}{N} \sigma_x\right)^2}$$
$$= \frac{\sigma_x}{\sqrt{N}} \quad \text{SDOM}$$

(what we see in reporting error)

---

### 3) Brownian Motion

- say we put 25 micron sphere in deionized water at room temp

- what would we see? (video)

" " " " with 40x mag.? (still video)

- what would we see if made movies with 40x digital microscope

(still video)

(depends on frame rate)

- lets make frame rate 0.25 fps <sup>2</sup> (ask for what the news)

(show movie)

391  
~~16~~  
 L3.6

we note: (ask for this)

- spheres appear to move
- "motion" is called Bragg-Motiv
- there may be other factors affecting their motion (query)

Why do this?

$$\overline{\Delta x^2} = 2 D_{ax} \Delta t$$

(also  $\overline{\Delta y^2} = 2 D_{ay} \Delta t$   
 $\overline{\Delta r^2} = 4 D_{ar} \Delta t$ )

- first, why  $\underline{\Delta x}$  ( $\underline{\Delta y}$ , or)?

- next what does  $\overline{\Delta x^2}$  mean?

- note that  $\sigma_{\Delta x}^2 = \overline{\Delta x^2} - (\overline{\Delta x})^2$

what do expect  $\overline{\Delta x}$  to be for BM?

[mention it should be 0, but this might be 'experiment zero' =>

$$\text{thus } \overline{\Delta x^2} \approx \sigma_{\Delta x}^2$$

- what is  $\sigma_{\Delta x}$ ? could we vary it?  
 how is that useful?

391  
~~391~~  
L3.7

Q1 m question, what is  $D$ ?  
(the 'diffusion-constant' first  
hypothesized by Einstein

with  $Df = k_B T$   $T = ?$ , units?

and  $f = 6\pi\eta R$  (drag factor, calculable)

$\Rightarrow$  If we can estimate  $D$  + its error ( $\delta D$ )  $\Leftarrow$   
then we can "  $k_B$  + its error

In this lab we will:

- 1) make movie of spheres undergoing BM
- 2) analyze there, first using Logger Pro
- 3) get LP data  $\Rightarrow$  python (or MC)
- 4) ? calc  $\Delta x$ ,  $\overline{\Delta x^2}$  vs  $\overline{\Delta t^2}$
- 5) ? do statistics  $\Downarrow$  random (data quality)
- 6) ? est  $D$  +  $\delta D$
- 7) calc  $k_B$  +  $\delta k_B$
- 8) 'Time Evolution'

ask  
students  
to  
complete  
last