

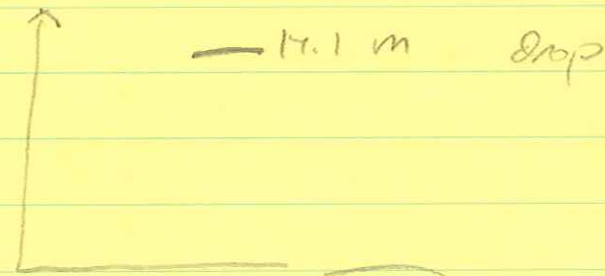
(L2.6)

2) Coding, in for loop for air resistance

Note, we didn't account for air resistance in previous problem?

⇒ could we fix this computationally? signs?
 need ^{new} ΔF on sked ball, $\equiv \vec{F}_g + \vec{F}_{air}$

$D = 2e^{-4}$
 (0.0002)



eqn $\frac{F_{tot}}{m} = a = \frac{v_f - v_i}{\Delta t}$

$v_f - v_i = \left(\frac{F_{tot}}{m}\right) \Delta t$
 $v_f = v_i + \left(\frac{F_{tot}}{m}\right) \Delta t$

$v = v_f$
 $v_f = \frac{\Delta v}{\Delta t}$
 $\Delta v = \frac{v_f - v_i}{\Delta t}$
 $v_{avg} = \frac{v_f + v_i}{2}$

$g_e = -7.8$ (trial)
 $y(1) = 14.1$
 $t(1) = 0$
 $v_i = 0$
 $\Delta t = 0.1$ (s)
 $F_{air} = g_e$ (trial)
 $v(i) = v_i + F_{air} * \Delta t$
 $v_{avg} = 0.5 * v(i)$
 $y(2) = y(1) + v_{avg} * \Delta t$
 for $i = 2 \dots 1000$

what is the F_{air} ??
 $F_a = \frac{F_{tot}}{m} = g_e - \frac{F_0}{m}$
 $= g_e - D v_{avg}^2$
 $= g_e - D v_{avg}^2$

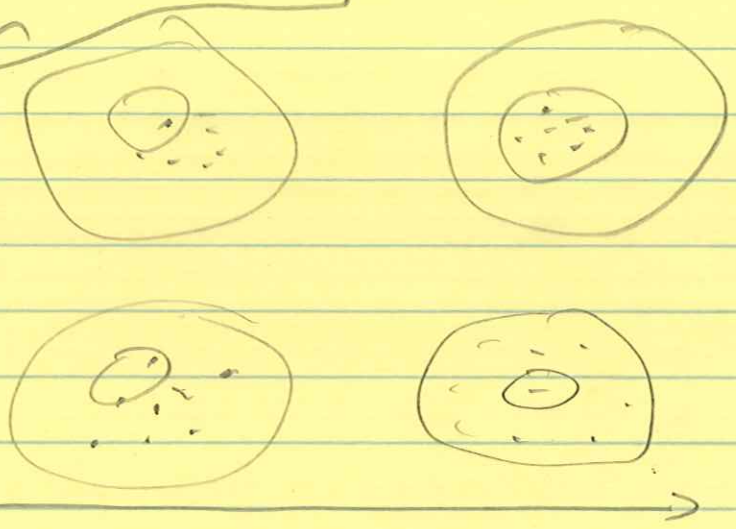
$F_0 = + D * v_{avg}^2 (i-1)^2$;
 $F_{air} = g_e + F_0$;
 $v(i) = v_i + (F_{air}/m) * \Delta t$;
 $v_{avg}(i) = (v_i + v(i))/2$;
 $v_i = v(i)$;
 $y(i) = y(i-1) + v_{avg}(i) * \Delta t$;

$\frac{h_f - h_i}{\Delta t} = v_{avg}$

- 1) more about error
- 2) python

321
w6
L2.1

(precision)



(accuracy)

(note student to suggest labels for plot above)

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

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

⇒ Born experiment [last up] $[125-127 \frac{GeV}{c^2}]$ actually

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

⇒ give example

take data until change

→ $125.3^{+0.6}$ CMS
and $126.0^{+0.6}$ ATLAS
(GeV/c)

391
u6
L2.1b

(example of reducing σ)

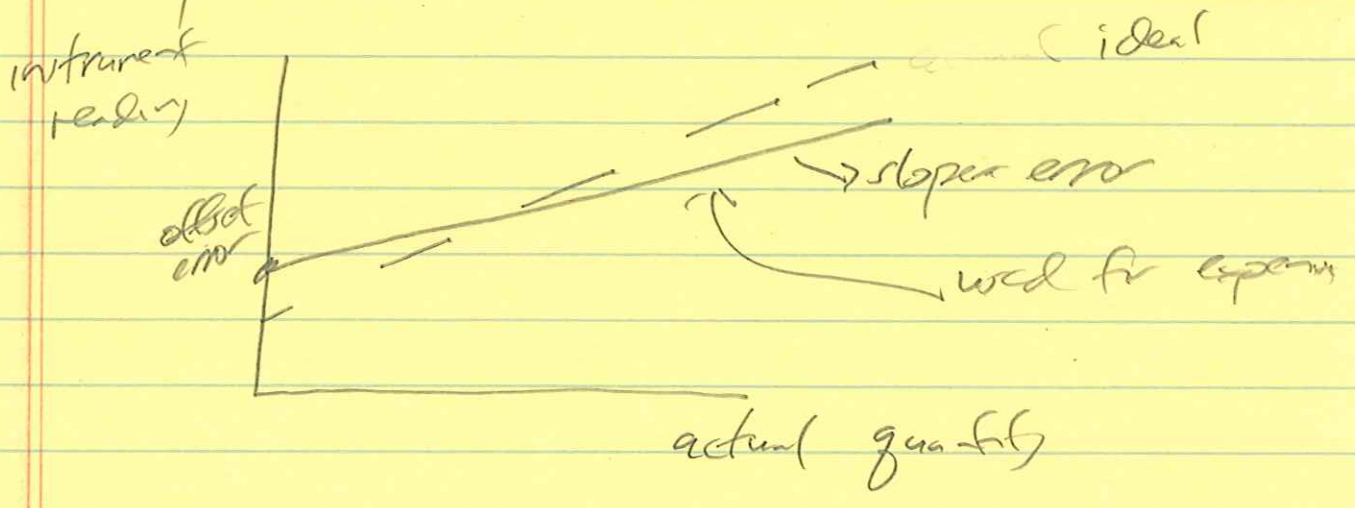
Higgs Boson, took ⁴⁰/₄ year
took ^{enough} data to where probability of getting
at least as strong a result
(for alt expl) was 1:3 million
and significance of 5 sigma (σ)

results were * $(125.3 \pm 0.6) \text{ GeV/c}^2$ CMS
 $(126.4 \pm 0.6) \text{ GeV/c}^2$ ATLAS

teams were blinded from each other since 2011

example of systematic error:

example would be mis-calibrated instrument



39
2016
L2.2

first will look at 1) increasing precision

a) consider distribution of ^{10 data} measurements

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

want to look at the distribution, so we

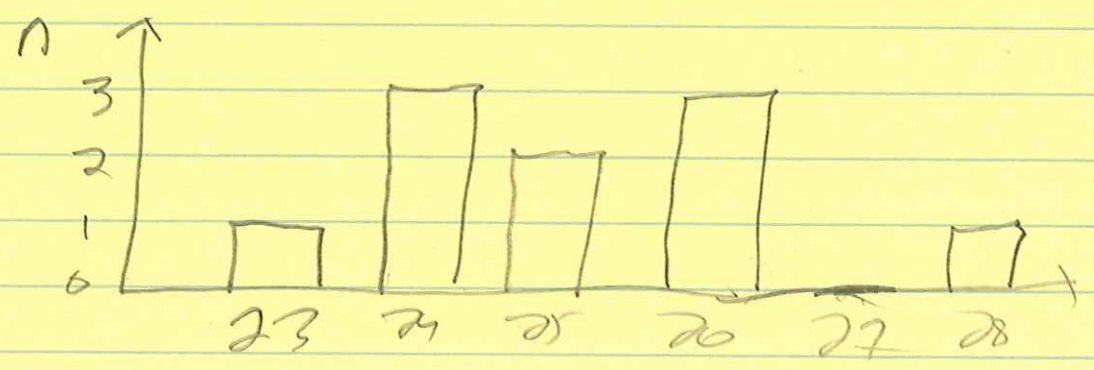
i) sort out the "data"

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

ii) count

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

iii) plot frequency of occurrence



371
W16
L2.3

iv) represent as mean

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

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

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

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

b) wide about real world "bin widths"

normally measure real number with some accuracy, say 0.1 and

distribution of 50 measure looks like the 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

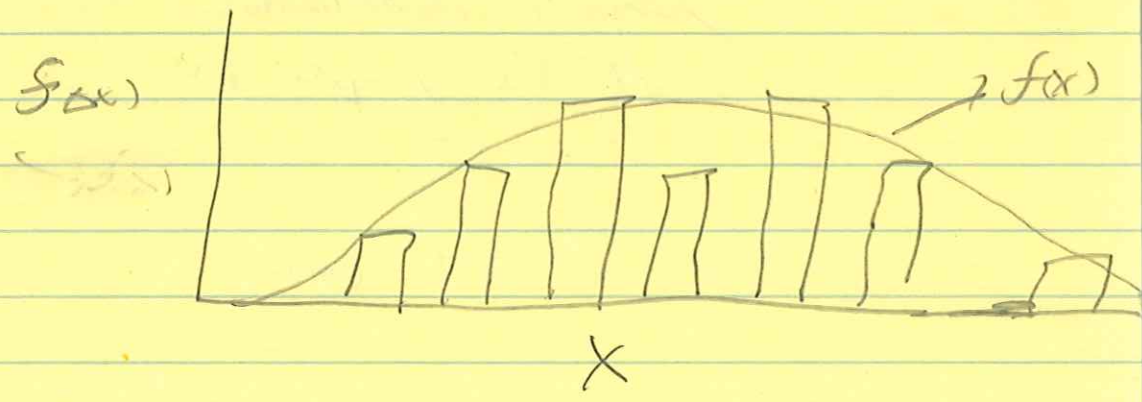
pick some reasonable average of gaps or "bin widths"

3π  
WLB  
L24

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

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

so will come from



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

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

or more properly formally

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

d) need to normalize s.t

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

381  
W16  
L2.5

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

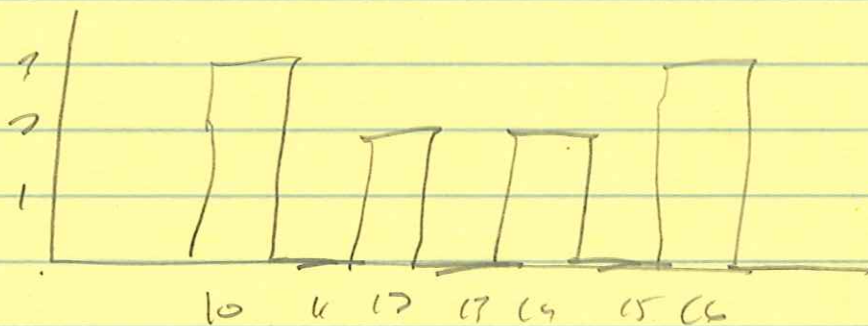
i)  $\bar{x} = \int_{-\infty}^{\infty} x f(x) dx$  "expectation value"

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

these two things define our prob distribution somewhat well

f) Central Limit Thm

Start with any distribution with well-defined mean and variance



can set a sample size  $N$

and take  $N$  samples (examples) 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  
W6  
L2.6

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.

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

it might not, itself, be "normal"

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

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



311  
W16  
L2.7

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  $\mu - b$  to  $\mu + b$  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