

391  
SP16  
L3.1  
E17

1) Normal distributions - PDFs & the Gaussian

2) SDOM - Standard deviation of the mean

3) ~~weighted averages~~

3A) Lab 2 Brownian Motion or  
Counting Statistics

- 1) typically ~~with~~ <sup>plan for</sup>  $N$  measurements of  
same thing for an experiment.  
(then we might change a 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 meas.  $x_i$  has  
a well-defined mean & standard devia-  
 $\bar{x} \pm \sigma$ .

Thus the central limit theorem says that  $M$  sets  
of these  $N$  measurings would be "normal". →

Look at stationarity + 2 results

39  
 F(7)  
 20  
 L3.2

first we'll look at i) increasing precision

a) consider distribution of  $(\pm 1)$  measurement

~~Details~~ 26, 29, 26, 28, 27, 29, 25, 26, 26, 25

want to look at the distribution, so we

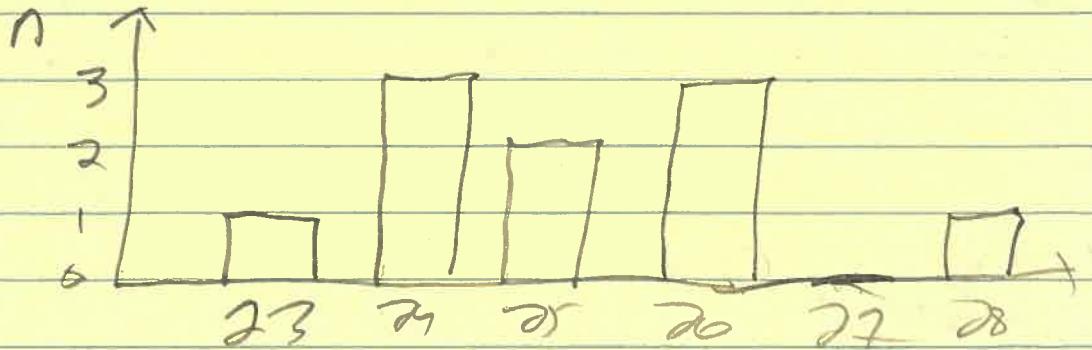
i) sort

23, 29, 29, 29, 25, 27, 26, 26, 26, 27, 28

ii) count

23	29	25	26	27	28
1	3	2	3	0	1

iii) pbf frequency of outcome



$\int f(x) dx$   
 $F(x)$   
 3.3

iv) represent as mean

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

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

v) code with  $f_j = \frac{n_j}{N}$

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

b) wide doesn't real world "bin widths"

normally measure real numbers with some accuracy, say 0.1 and

so measure looks instead like the  
[already ordered]

$$23.7, 23.8, 23.9, 23.9, 24.6, \dots$$

$\sim$      $\sim$      $\sim$      $\sim$   
 0.1    0.1    0.0    0.7

look

at

space  
in data

pick some reasonable average of gaps  
as "bin widths"

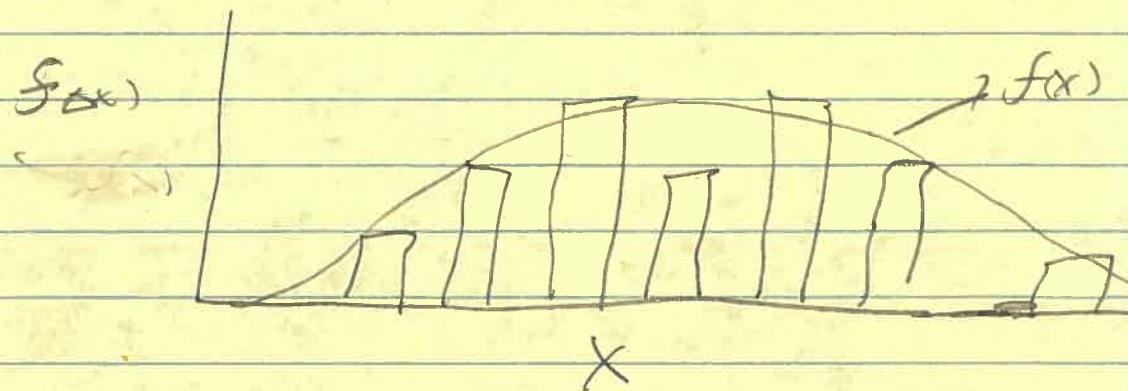
3x

(1-17)k  
L273.4

(1) we're working towards a  
statistical description of data distribution

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

so will move from



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

→ Then  $f(x) dx$  is the <sup>fraction</sup> of samples between  $x$  and  $x + \Delta x$

or more properly formally

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

d) need to normalize S.F.

$$\int f(x) dx = 1$$

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

these two things define our distribution well

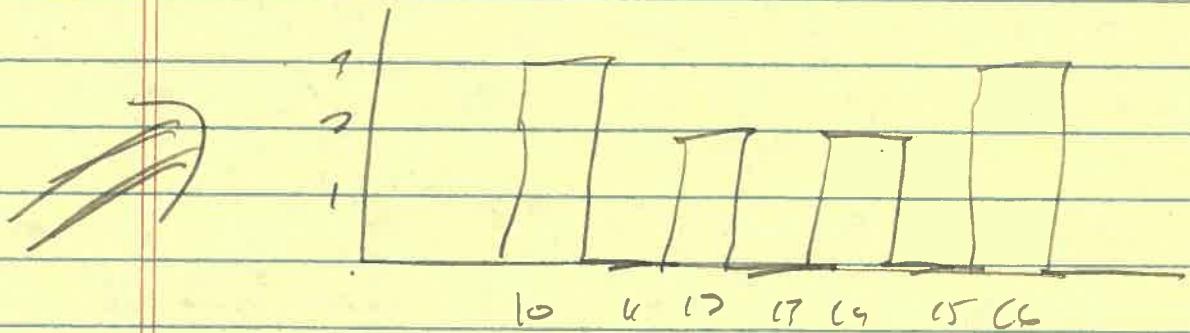
$$\text{i) } \bar{x} = \int_{-\infty}^{\infty} x f(x) dx \quad \begin{matrix} \leftarrow \text{expectation} \\ \text{value} \end{matrix}$$

$$\text{ii) } \sigma^2 = \int_{-\infty}^{\infty} (x - \bar{x})^2 f(x) dx \quad \begin{matrix} \leftarrow \text{variance} \end{matrix}$$

f) Central Limit Thm

skip

Start with any distribution with well-defined mean and variance



Can set a sample size  $N$

and take  $N$  samples (examples) of above

$$\text{say } N=4 \Rightarrow$$

$$\{10, 10, 14, 16\} \Rightarrow 12.5$$

Keep doing this, then plot freq. dist of sample

get something that looks like a "normal distribution"

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

CLT says that if  $N \rightarrow \infty$   
freq. distn.  $\rightarrow$  "Normal"  
with well-defined mean and variance.

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

~~important~~  $\rightarrow$  it might not, itself, be "normal"

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

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

$$\text{or } e^{\frac{(x-\bar{x})^2}{2\sigma^2}}$$

3.1

F7

L3.7

for now we'll call  $f(x)$  the "width fn"  
and  $\bar{x}$  the "maximally probable value"

→ we'll typically need to normalize this  
fx) s.t.

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

### b.) ERF

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

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

$$\text{Can let } z \equiv \frac{x-\mu}{\sigma} \Rightarrow$$

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

called the "error fn" or ERF

look up when in form

2) SDom

Just mention

additional  
assumptiontake the  
true value  
of  $\bar{x}$   
represented  
by  $x_i$ have established, given certain (reasonable)  
assumptions,  $\bar{X} = \bar{x}$  and  $\sigma_x = \delta_x$ Still working w/  $N$   $x_i$ , what is "reliability"  
of  $\bar{X} = \bar{x}$ ?

to consider, look at  $\bar{X}_k = \frac{\sum x_{i,k}}{N}$   $k=1, m$   
 dist. of  $\bar{x}_k$  over  $M$  "sets"

we note that if  $x_1, x_2, \dots, x_N$  is normally distributed  
 then  $\bar{x}$  is normally distributed (sample mean of  $x_i$ )

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

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

Now look at standard deviation as result

So measuring  $M$  sets doesn't change  $\bar{X} = \bar{x}$  (if  $\delta_x$  is small)

(Q) What's width estimate of  $\bar{X}$ ?

add error in  $x$   
 in quadrature

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

but  $\delta_{x_1} = \delta_{x_2} = \dots = \delta_x$

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

$$\Rightarrow \delta_{\bar{x}} = \sqrt{\left(\frac{1}{N} \delta_x\right)^2 + \dots} = \left(\frac{1}{N} \delta_x\right)$$

$$= \frac{\delta_x}{\sqrt{N}} \quad \text{SDOM}$$

(What we see in reporting error)

### 3) Brownian Motion

- Say we put ~5 micron sphere in deionized water at room temp

- what would we see? (a-d)

" " " " with 40x mag? (still a-d)

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

(still a-d)  
(depends on frame rate)

- lets make frame rate 0.05 fps (ask for what)  
the mean

Show movie)

we note: (ask for this)

- planets appear to move
- "motion" is called Branca-Motion
- there may be other factors affecting their motion (Zver)

Why do this?

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

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

- first, why  $\Delta x$  ( $\Delta y$ , 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 B.M?

[mention it should be  $\phi$  - but this might be 'experimental zero'  $\Rightarrow$   $\overline{(\Delta x)^2} \ll \overline{\Delta x^2}$ ]

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

- what is  $\Delta t$ ? Could we vary it?  
 how is that useful?

3gr  
F17  
L3.7

• #1 m question what is  $D$ ?

(the 'diffusion-constant' first hypothesized by Einstein)

with  $D = k_B T$

$T = ?$ , units?

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

⇒ If we can estimate  $D$  & its error ( $\delta D$ )  
then we can " "  $k_B$  & its error

In this lab we will:

1) make movies of spheres undergoing Brownian motion

2) analyze these, first using Logger Pro

3) get  $L^2$  data  $\Rightarrow$  python (or MC)

ask  
student  
to  
complete  
last

4) ? calc  $\Delta s$ ,  $\overline{\Delta x}^2$  vs  $\overline{\Delta t}^2$

5) ? do statistics  $P_{\text{random}}$  (data quality)

6) ? est  $D + \delta D$

7) calc  $k_B + \delta k_B$

8) 'Time Evolution'