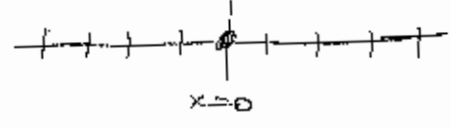


RANDOM WALK

- MODEL FOR PARTICLE(S) MOVING RANDOMLY THRU FLUID (DIFFUSION) (KEIF gives other applications; tools are general)

1 DIM:

PARTICLE STARTS AT $x=0$



ASSUMPTIONS:

- ① GOES DISTANCE l BEFORE HITTING NEXT PARTICLE (MEAN FREE PATH); REPEATS
- ② AFTER COLLISION, DIR. OF NEXT STEP INDEP. OF PREVIOUS ST
 PROB NEXT STEP TO RT: p ($0 < p < 1$)
 " " " TO LT: $q = 1 - p$

IF (HOMOG: $p = q = \frac{1}{2}$; IF $p > q \sim$ FLUID FLOWS TO RT)
 (treating w/ probability, since don't want to get into details of each collision)

PROBLEM:

AFTER N STEPS, FIND PROB. $P(m)$ FOUND AT $x = ml$
 (m AN INTEGER)

{ CLASSIC FORMULATION : DRUNK AT LAMP POST ON SIDEWALK }
{ says $p=q=1/2$ in adv: $x=0$ but dist. $x \neq 0$ }

PROBABILITY: IN ^{LARGE} ENSEMBLE OF IDENT SYSTEMS, AFTER N STEPS
 $P(m) = (\# \text{ AT } ml) / N_{\text{ENS}}$

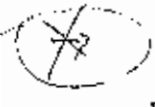
REALIZATIONS:

- (a) 1 PART. EACH IN N_{ENS} FLUIDS
- (b) " " N_{ENS} TIMES IN 1 " " \Rightarrow DIFFUSION OF DYE
- * (c) MANY PARTICLES IN 1 " " \Rightarrow $P \propto$ DENSITY CAN SEE DIRECTLY

{ REIF: PROB. DEPENDS ON ENSEMBLE

ex prob seed yields red flower - diff if seed \in {tulips}
 vs seed \in {all plants, incl trees}

\Rightarrow prob given the info or constraints in ensemble?



skip

$p = q = \frac{1}{2}$

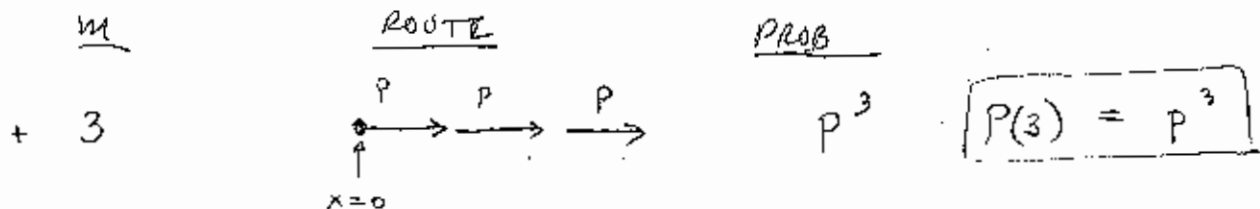
EXPECT:

- WILL USUALLY END CLOSE TO $x=0$ (AND $\bar{x}=0$ BY SYMM.)
- RARE FOR $m = \pm N$ (ESP FOR N LARGE) \Rightarrow 1 WAY; BUT LOTS OF WAYS TO GET TO $x \sim 0$
- AVE DISTANCE $\neq 0$

$p > q$: - ON AVE, TAKE MORE RT STEPS THAN LEFT \Rightarrow MIGRATES TO RT

- CAN GUESS: $\bar{x} = N(p-q)$ l.
(will confirm)

ex $N=3$ (start simple)

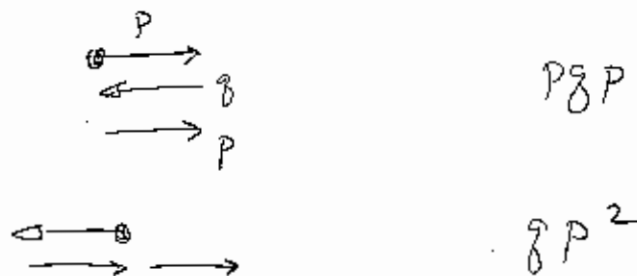
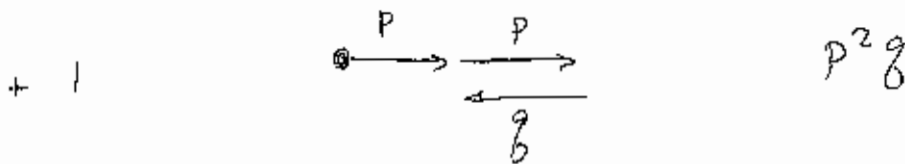


just remind use both here

INDEPENDENT
TWO EVENTS A, B w/ PROB $P(A)$, $P(B)$
 PROB BOTH HAPPEN: $P(A) \cdot P(B)$ ($\leq P(A)$ OR $P(B)$)
 " EITHER " : $P(A) + P(B)$ ($\geq P(A)$ OR $P(B)$)
INDEP: VALUE GET FOR A HAS NO EFFECT ON B

ex 2 coins: prob of 2 heads $\frac{1}{2} \times \frac{1}{2} = \frac{1}{4}$
 prob both same $\frac{1}{2} + \frac{1}{2} = \frac{1}{2}$

+ 2 CAN'T DO IT 0 { IF N ODD M MUST BE ODD }



TOTAL:
 $P(1) = 3p^2 q$

<u>m</u>	<u>ROUTE</u>	<u>PROB</u>
0	CAN'T	0

-1 ETC → JUST REPLACE $p \leftrightarrow q \Rightarrow P(-1) = 3 p q^2$

-3 q^3 $P(-3) = q^3$

$$\left\{ \begin{array}{l} \text{ex } p=q=\frac{1}{2}: \\ P(3) = P(-3) = \frac{1}{8} \\ P(1) = P(-1) = \frac{3}{8} \end{array} \right\}$$

HAVE ALL CASES.

PROB THAT PARTICLE GOES ANYWHERE? 1

CHECK: $\sum_{m=-3}^3 P(m) = 1$

$$\Rightarrow p^3 + 3p^2q + 3pq^2 + q^3 = (p+q)^3 = 1^3 = 1 \quad (\text{any } p, q)$$

IN GENERAL:

$n_R, n_L = \#$ RT, LT STEPS

$$\begin{aligned} N &= n_R + n_L \\ m &= n_R - n_L = 2n_R - N \end{aligned}$$

OR

$$\begin{aligned} n_R &= \frac{1}{2}(N+m) \\ n_L &= \frac{1}{2}(N-m) \end{aligned}$$

put on side board

TYPICAL ROUTE TO 'mL':

STEP 1 2 3
R R L R LL ... R

$n_{R,L} = 0, 1, 2, \dots, N$
 N, m BOTH EVEN OR BOTH ODD
 $\left. \begin{array}{l} n_R \\ n_L \end{array} \right\} \begin{array}{l} R's \\ L's \end{array}$

PROB FOR THIS SPECIFIC ROUTE

$$p^{n_R} q^{n_L} = p^{\frac{1}{2}(N+m)} (1-p)^{\frac{1}{2}(N-m)}$$

(note: if N, m fixed, so are n_R, n_L)

HOW MANY WITH THIS n_R & n_L ?

\Rightarrow # WAYS TO ^{ARRANGE} n_R R'S & n_L L'S IN N SPACES/STEPS:

(1) ASSUME HAVE N DISTINCT OBJECTS

		STEP	1	2	3	...	N
1st	:						N CHOICES
2nd	:		$N-1$	"		ETC	{ for each one of the N choices for 1st

$\Rightarrow N!$ WAYS TO ARRANGE

(2) WE OVER COUNTED:

NOW RECOGNIZE THAT R'S SAME, L'S SAME

R R L R L L L R R

SAME IF REARRANGE R'S AMONG SELVES; ALSO L'S

$n_R!$ WAYS TO ARRANGE R'S (counting same way)

$n_L!$ WAYS TO " L'S

\Rightarrow NUMBER OF DISTINCT ARRANGEMENTS WITH n_R R'S & n_L L'S

$$\frac{N!}{n_L! n_R!} = \frac{N!}{(N-n_R)! n_R!} \equiv \binom{N}{n_R}$$

"BINOMIAL COEFFICIENT"

$$= \frac{N!}{\left[\frac{1}{2}(N+m)\right]! \left[\frac{1}{2}(N-m)\right]!}$$

NOTE:

CS 1.5

BINOMIAL EXPANSION:

$$(x + y)^N = \binom{N}{0} x^N + \binom{N}{1} x^{N-1} y^1 + \binom{N}{2} x^{N-2} y^2 + \dots + \binom{N}{N-1} x y^{N-1} + \binom{N}{N} y^N$$

$$= \sum_{n=0}^N \binom{N}{n} x^{N-n} y^n$$

{ 0! = 1 }

WHY?

ex $(x + y)^3 = (x + y)(x + y)(x + y)$

$$= x^3 + \underbrace{(x^2y + xyx + yx^2)}_{3x^2y} + \dots$$

ALL POSSIBLE TERMS w/
2 x's AND 1 y : $xx y + x y x + \dots$

SAME COUNTING

FINALLY

$$P(m) = \frac{N!}{\left[\frac{1}{2}(N+m)\right]! \left[\frac{1}{2}(N-m)\right]!} p^{\frac{1}{2}(N+m)} q^{\frac{1}{2}(N-m)}$$

$q = 1 - p$
 $m \leq N$ BOTH ODD OR BOTH EVEN; ELSE 0

BINOMIAL DISTRIBUTION

CHECK: q SMALL: $P(m)$ SMALL UNLESS $N \sim +m$

BINOM. COEFFS LARGEST WHEN $m \sim 0$

⇒ PICTURES

$n_R \sim n_L \sim N/2$

* CAN ALSO EXPRESS AS PROB. OF n_R RT. STEPS:

$$W(n_R) \equiv P(m) = \frac{N!}{n_R! (N - n_R)!} p^{n_R} q^{(N - n_R)}$$

(some info)



$$\sum_{m=-N}^N P(m) \stackrel{?}{=} 1$$

EASIEST IF USE FORM

$$W(n_R) = \frac{N!}{(N-n_R)! n_R!} p^{n_R} q^{N-n_R}$$

don't re-write if don't have to

skip { (WITH $q = 1 - p$ $n_R + n_L = N$
 $n_R - n_L = m$)

$$\sum_{n_R=0}^N W(n_R) = \sum_{n_R=0}^N \binom{N}{n_R} p^{n_R} q^{N-n_R}$$

$$= (p + q)^N = 1$$



skip { WILL SEE SEVERAL TRICKS FOR SUMMING SERIES.
cf GRADSHTEYN & RYZHIK; SYMBOLIC PROGRAMS ~ MATHEMATICA
MAPLE }

skip ex USEFUL FOR PROB 1.5.C (cf A.1)

$$f(x) = \sum_{n=0}^N x^n = 1 + x + x^2 + \dots + x^N$$

$$x f(x) = x + x^2 + \dots + x^N + x^{N+1}$$

SUBTR

$$(1-x) f(x) = 1 - x^{N+1}$$

$$f(x) = \frac{1 - x^{N+1}}{1 - x}$$

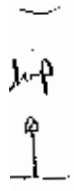
(converges if $|x| < 1$)
 $N = \infty \Rightarrow 1 + x + x^2 + \dots = \frac{1}{1-x}$
also $\frac{1}{1+x} = 1 - x + x^2 - x^3 + \dots$

SPECIAL CASE!

$$\sum_{n=0}^N e^{ny} = \sum_{n=0}^N (e^y)^n$$

\Rightarrow SAME AS ABOVE

$$= \frac{1 - e^{(N+1)y}}{1 - e^y}$$



$$N = 20$$

$$p = q = \frac{1}{2}$$

GENERAL DISCUSSION OF MEAN VALUES

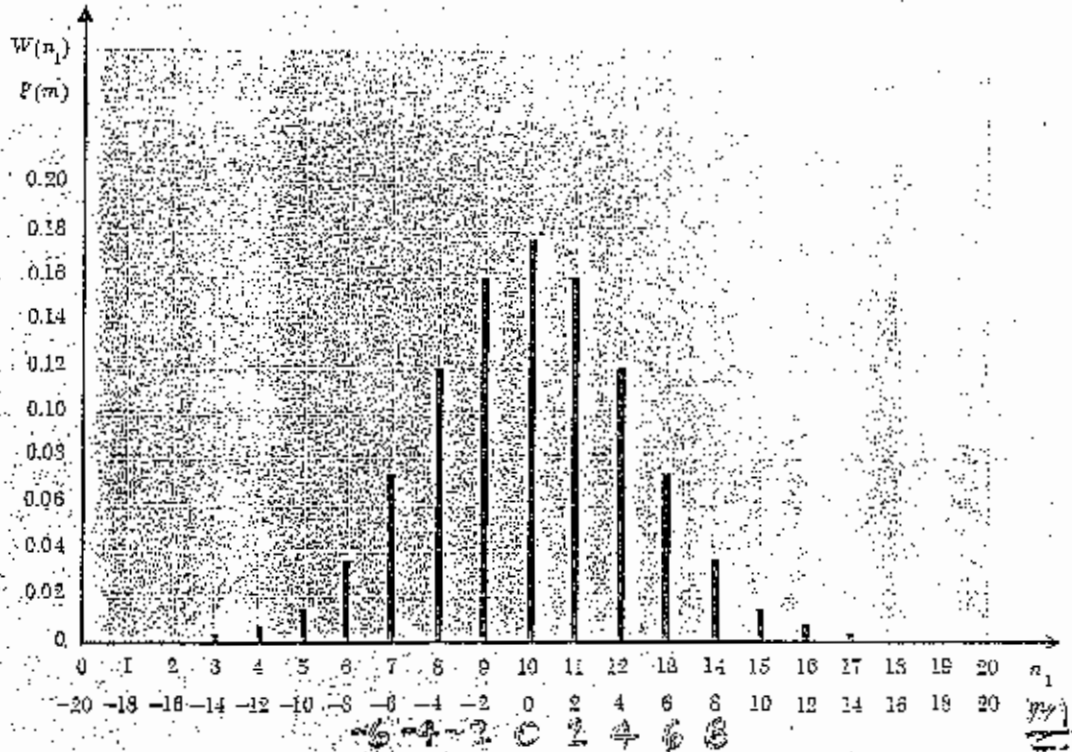
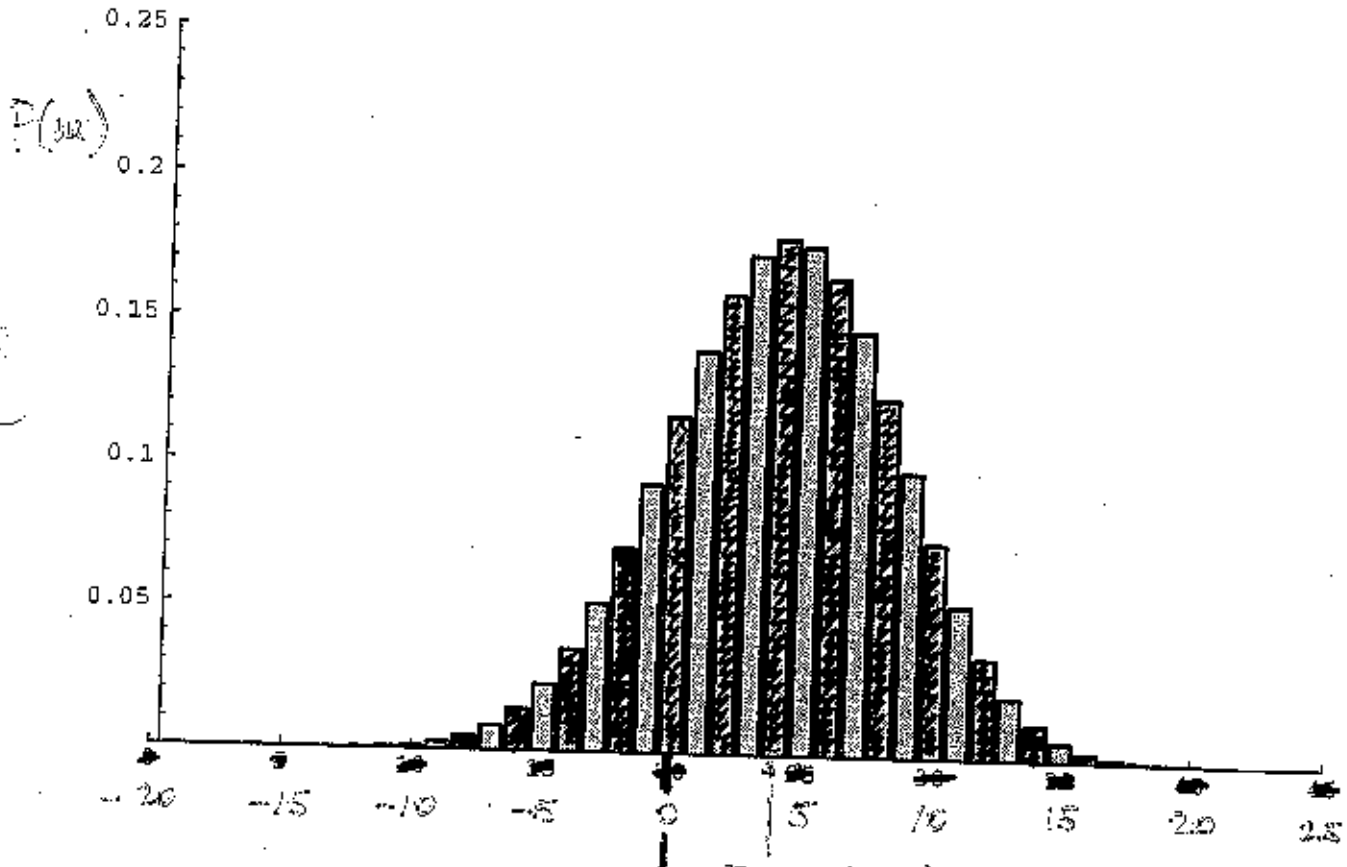


Fig. 1.2.3 Binomial probability distribution for $p = q = \frac{1}{2}$ when $N = 20$ steps. The graph shows the probability $W_N(n_1)$ of n_1 right steps, or equivalently the probability $P_N(m)$ of a net displacement of m units to the right.

$N = 20$

$p = .6 \quad q = .4$

(ONLY GOOD FOR N EVEN)



$\bar{M} = N(p - q) = 4$



$\sqrt{(\Delta M)^2} = \sqrt{4pqN}$

COMPUTE SOME AVES w/ RAND WALK:

(EASIEST TO WORK w/ $W(n_R) \Rightarrow$ DON'T HAVE TO WORRY ABOUT EVEN/ODD)

$$(1) \text{ HAVE } \sum_{n_R=0}^N W(n_R) = 1$$

$$(2) \bar{n}_R = \sum_{n_R} W(n_R) n_R = \sum \binom{N}{n_R} p^{n_R} q^{N-n_R} \cdot n_R$$

TRICK:

(a) TREAT p, q AS INDEP. ; i.e. $\bar{n}_R(p, q)$ } i.e. $q \neq 1-p$ yet

(b) NOTE $p \frac{\partial}{\partial p} p^{n_R} = n_R p^{n_R}$

$$(c) \bar{n}_R(p, q) = \sum_{n_R} \binom{N}{n_R} \left(p \frac{\partial}{\partial p} p^{n_R} \right) q^{N-n_R}$$

$$= p \frac{\partial}{\partial p} \left\{ \sum_{n_R} \binom{N}{n_R} p^{n_R} q^{N-n_R} \right\}$$

CAN NOW SUM:

$$(p+q)^N$$

DERIV: $p N (p+q)^{N-1}$

$$\Rightarrow \bar{n}_R(p, q) = \sum \binom{N}{n_R} p^{n_R} q^{N-n_R} n_R = p N (p+q)^{N-1}$$

n_R slip true for any p, q

(d) NOW TAKE $q = 1-p$

$$\bar{n}_R = \bar{n}_R(p, q) \Big|_{q=1-p} = Np$$

IN SAME WAY

$$\boxed{\bar{n}_R = Np}$$

$$\boxed{\bar{n}_L = Nq}$$

DID YOU ~~WANT~~ GUESS THIS?

$$\Rightarrow \boxed{\bar{n}_A = \bar{n}_R - \bar{n}_L = N(p-q)}$$

(w/ $q = 1-p$)

(3) SPREAD : (a VARIANCE)

$$\overline{(\Delta m)^2} = \overline{(m - \bar{m})^2} = \overline{m^2} - \bar{m}^2$$

\checkmark
 $\equiv (\Delta^* m)^2$

DO USING n_R , $W(n_R)$

$$m = n_R - n_L = 2n_R - N \quad (\text{on side lft})$$

$$\begin{aligned} \Delta m = m - \bar{m} &= (2n_R - N) - (2\bar{n}_R - \bar{N}) \\ &= 2(n_R - \bar{n}_R) = 2\Delta n_R \end{aligned}$$

$$\Rightarrow \overline{(\Delta m)^2} = 4 \overline{(\Delta n_R)^2} = 4 (\overline{n_R^2} - \bar{n}_R^2)$$

$\stackrel{\Delta}{=} N^2 p^2$

USE SAME TRICK:

$$\begin{aligned} \overline{n_R^2} &= \sum_{n_R} W(n_R) n_R^2 = \sum_{n_R} \binom{N}{n_R} p^{n_R} q^{N-n_R} n_R^2 \\ &= \left(p \frac{\partial}{\partial p}\right) \left(p \frac{\partial}{\partial p}\right) (p+q)^N \Big|_{q=1-p} \end{aligned}$$

(SOME ALGEBRA, AFTER $q \equiv 1-p$)

$$\overline{n_R^2} = N^2 p^2 + N p q$$

$$\overline{(\Delta m)^2} = 4 (\overline{n_R^2} - \bar{n}_R^2) = 4 N p q$$

$$\Delta^* m \equiv \sqrt{\overline{(\Delta m)^2}} = \sqrt{4 N p q}$$

(can see them
on previous
picture)

$$\rightarrow \sqrt{N} \quad \text{if } p=q=\frac{1}{2}$$

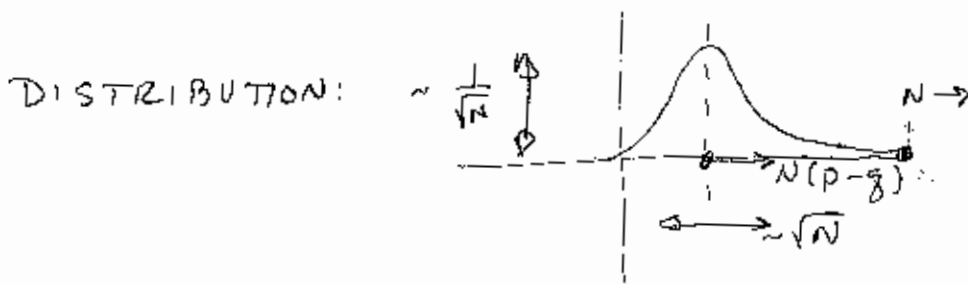
SOME GENERAL FEATURES:
vs N

ALREADY:
 $\Delta^* m = \sqrt{4Npq}$ $\Delta^* n_2 = \sqrt{Npq}$

1.12

PEAK (will see later)

$$P(m_{\text{MAX}}) \approx \frac{1}{(2\pi Npq)^{1/2}} \begin{cases} \text{skip} \\ \rightarrow \left(\frac{2}{\pi}\right)^{1/2} \left(\frac{1}{N}\right)^{1/2} \\ p=q=\frac{1}{2} \end{cases}$$



(picture)

SHRINKS $\frac{1}{2}$ SPREADS, BUT AREA CONST (why? prob dist)

same for moving plot

(WILL SEE IN HW, CAN THINK OF AS DESCRIBING t DEVELOPMENT OF DIFFUSION PROCESS)

(THERE $NL = vt$
 \uparrow
 over v)

IN TERMS OF n_R & $W(n_R)$:

$$\bar{n}_R = Np$$

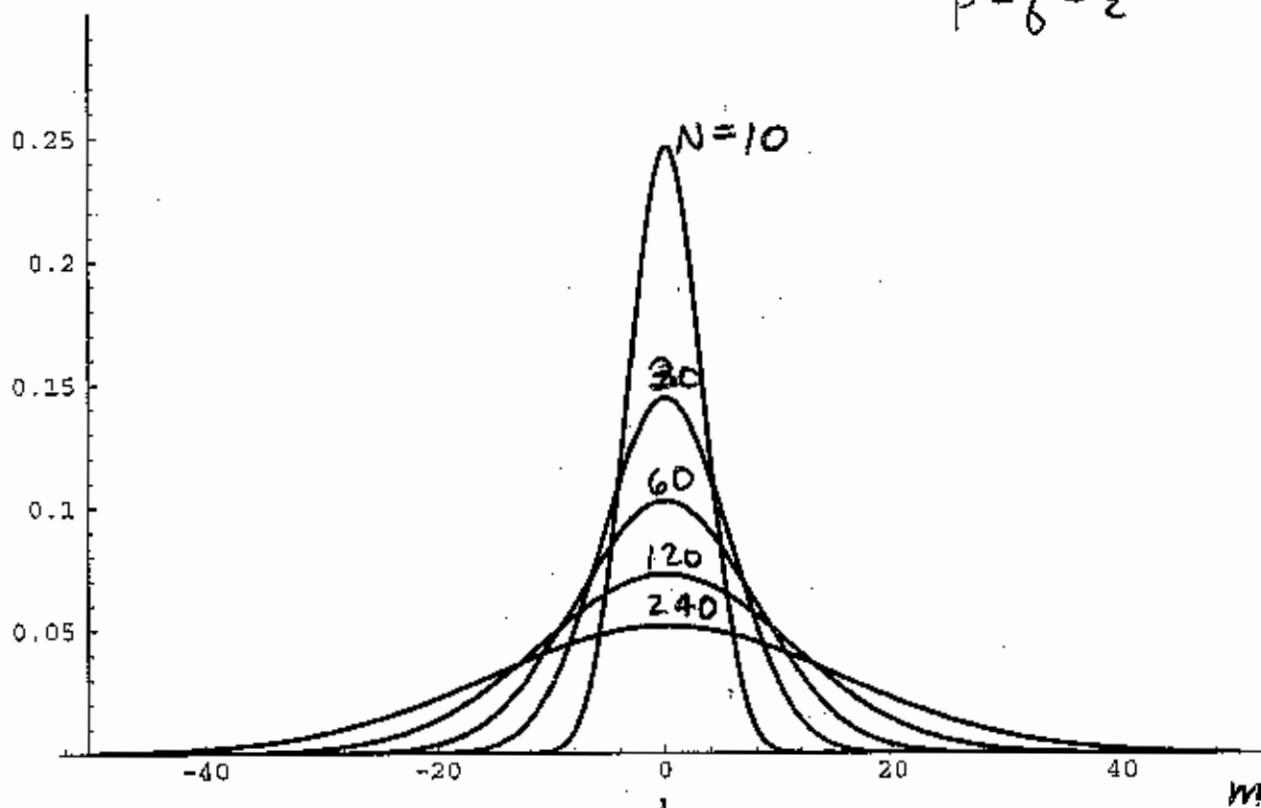
$$\Delta^* n_R = \frac{1}{2} \Delta^* m = \sqrt{Npq}$$

RELATIVE UNCERTAINTY: $\frac{\Delta^* n_R}{\bar{n}_R} = \frac{1}{\sqrt{N}} \left(\frac{q}{p}\right)^{1/2}$

SHRINKS $\cdot W/N \Rightarrow$ % ACCURACY INCREASES

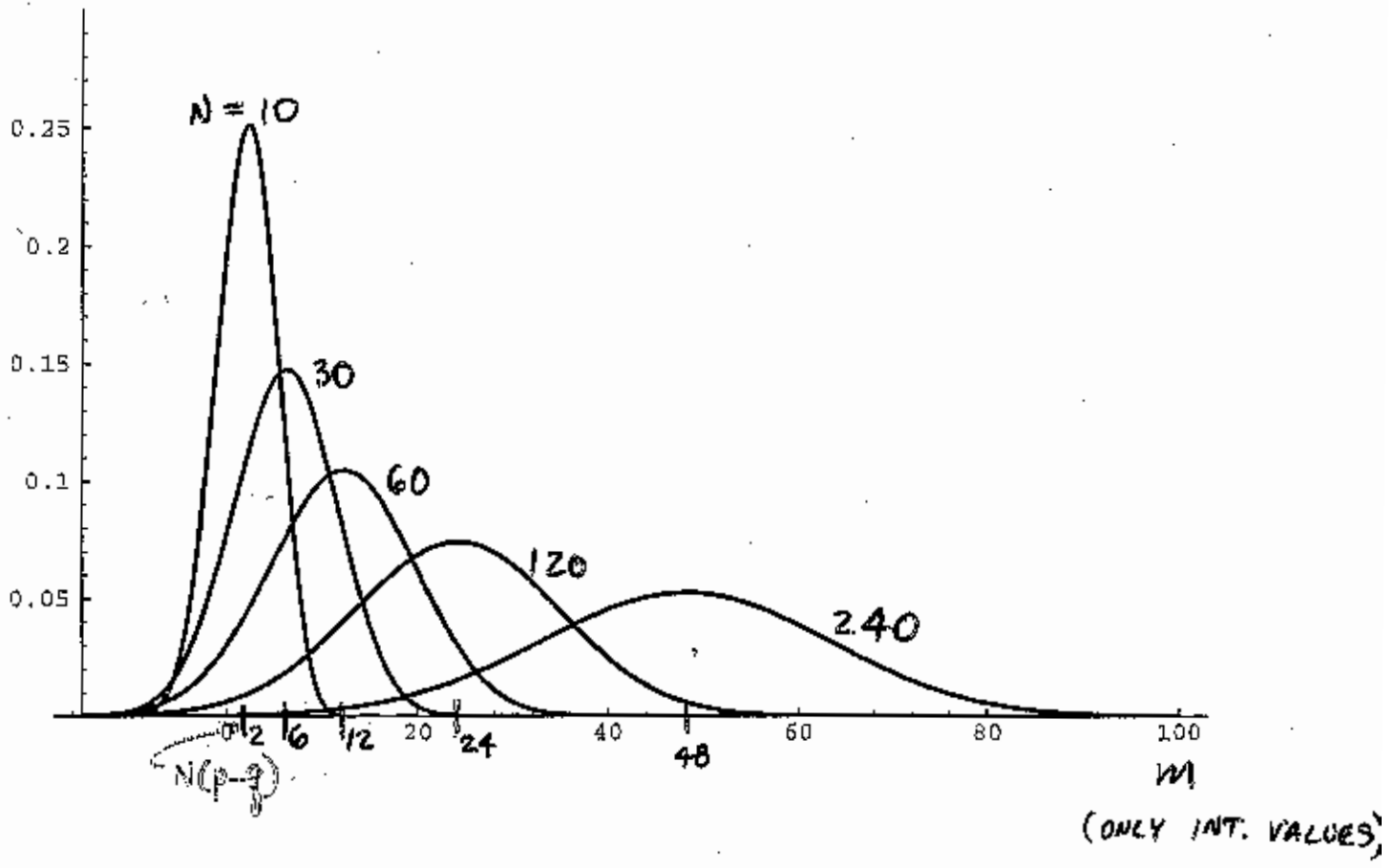
(ex $p=q=\frac{1}{2}$ \bar{n}_R MORE FOCUSED AT $N/2$)

$$p = q = \frac{1}{2}$$



- $\sqrt{10}$
- $\sqrt{30}$
- $\sqrt{60}$
- $\sqrt{120}$
- $\sqrt{240}$

$p = .6$ $q = .4$



CONTINUOUS APPROX

{ will work very hard to get only an approx soln from an exact soln

C.I.

USEFUL FOR N LARGE

ex MOLECULES IN FLUID:

$x \sim$ MACRO DIST ~ 1 cm

$l \sim$ ATOMIC $\sim 1 \text{ \AA} \sim 10^{-10} \text{ m} \ll x$

$$\bar{x} = \bar{v} l = N(p-q)l \Rightarrow N \sim 10^8 \text{ COLLISIONS BEFORE PEAK MOVES 1cm}$$

(A)

INTERESTING LIMIT:

$$\bar{x} \sim N l \begin{matrix} \leftarrow \text{SMALL} \\ \uparrow \\ \downarrow \text{LARGE} \end{matrix}$$

REASONABLE (w MACRO)

NOTE:

$P(m)$ (N=240 PLOT)

(1) $P(m) \rightarrow$ SMOOTH (w $\Delta P \ll P$) { jumps small rel. to what? }

(2) FOR MACRO MEAS, UNCERTAINTY IN x IS $\gg l$:

\Rightarrow AVE. SOME x MEASUREMENTS: SOMETIMES FEW MORE, SOMETIMES LESS l 's

\Rightarrow MORE LIKELY MEAS. DENSITY IN $x \Rightarrow$ ALSO " " \Rightarrow SMOOTHS OUT

\Rightarrow UNCERT. IN $p \gg$ JUMPS

\Rightarrow CONTINUOUS APPROX. SHOULD WORK OK

(also, this is a model; if meas. are acc. enough to see atoms, better treat it microscopically if more detailed approx; contin approx should be as good as discrete model)

REIF NOTATION:

$dx =$ INFINITESIMAL BUT MACROSCOPIC

$\delta x =$ SHORTEST (ATOMIC) DIST.

USUALLY $x \gg dx \gg \delta x$

(limit $dx \rightarrow 0$ really means make it small enough that $f(x)$ doesn't change much over dx)

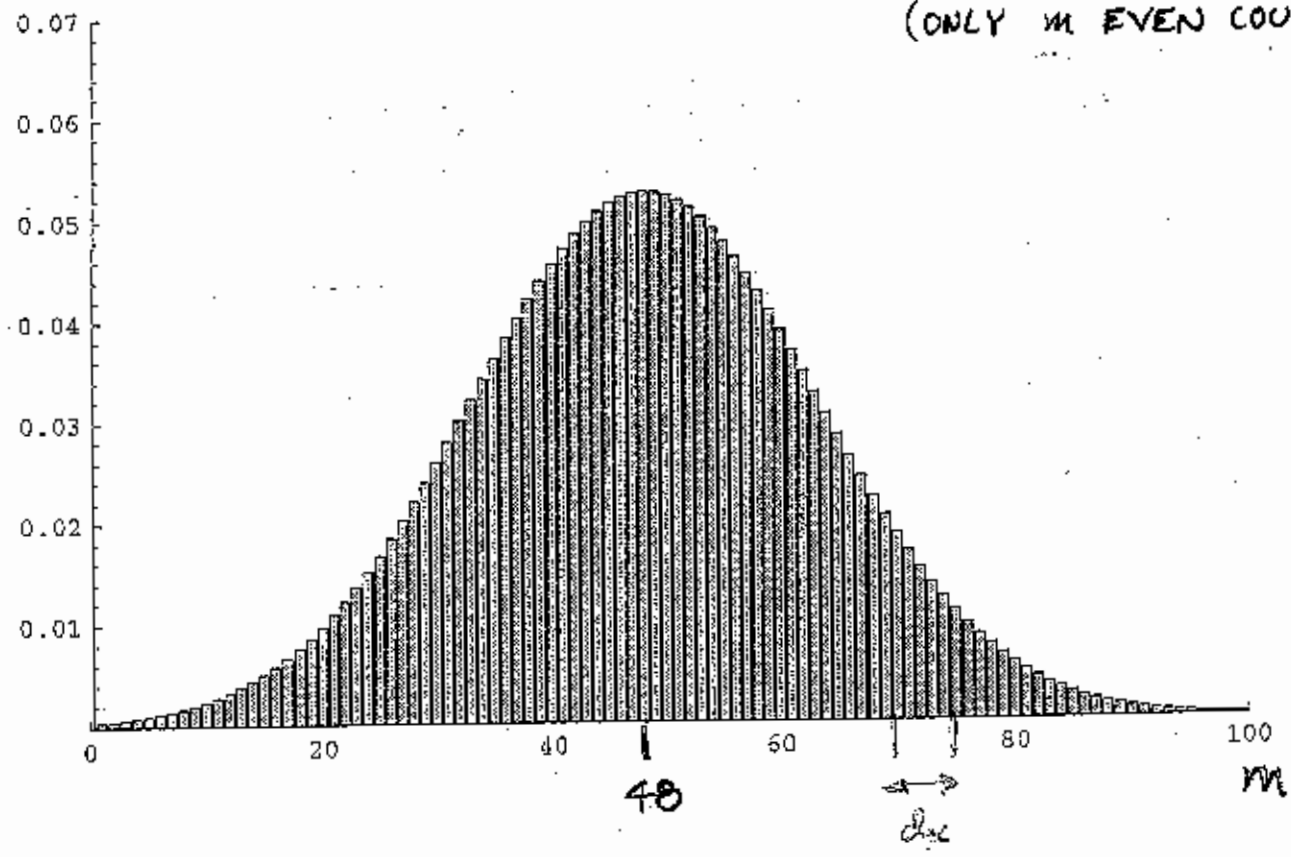
should only have
values at 2's, a
see $W(n, x)$

$P(m)$

$N = 240$

$p = .6$ $q = .4$

(ONLY m EVEN COUNT)



DEFINE CONTINUOUS DISTR:

$$x_m \equiv ml \quad m = -N, -N+2, \dots, N-2, N$$

(recall m even or odd)

PROB DENSITY

$$P(x_m) = \lim_{l \rightarrow 0} \frac{P(m)}{2l}$$

SMALLEST Δx

PROBLEM:

$P(m)$ HAS FACTORIALS:

- AWKWARD TO USE AS $N \rightarrow \infty$
- ONLY DEF'D FOR INTEGER VALUES:
- $l \rightarrow 0$ LIMIT PAINFUL

SOL'N:

REPLACE $P(m)$ w/ ^{SIMPLE} "CONTINUOUS APPROX":

WILL SEE:

EASY TO \int

CAN READ OFF \bar{x} , Δ^*x

UNIVERSAL \Rightarrow GOOD APPROX FOR ^{ALMOST} ANY LARGE SUM OF RANDOM VARS

\Rightarrow GAUSSIAN



PROBLEM:

$P(m)$ BECOMES SMOOTH BUT SHARPLY PEAKED:

$$\Delta^*x \sim N^{1/2}l \quad \bar{x} \sim Nl$$

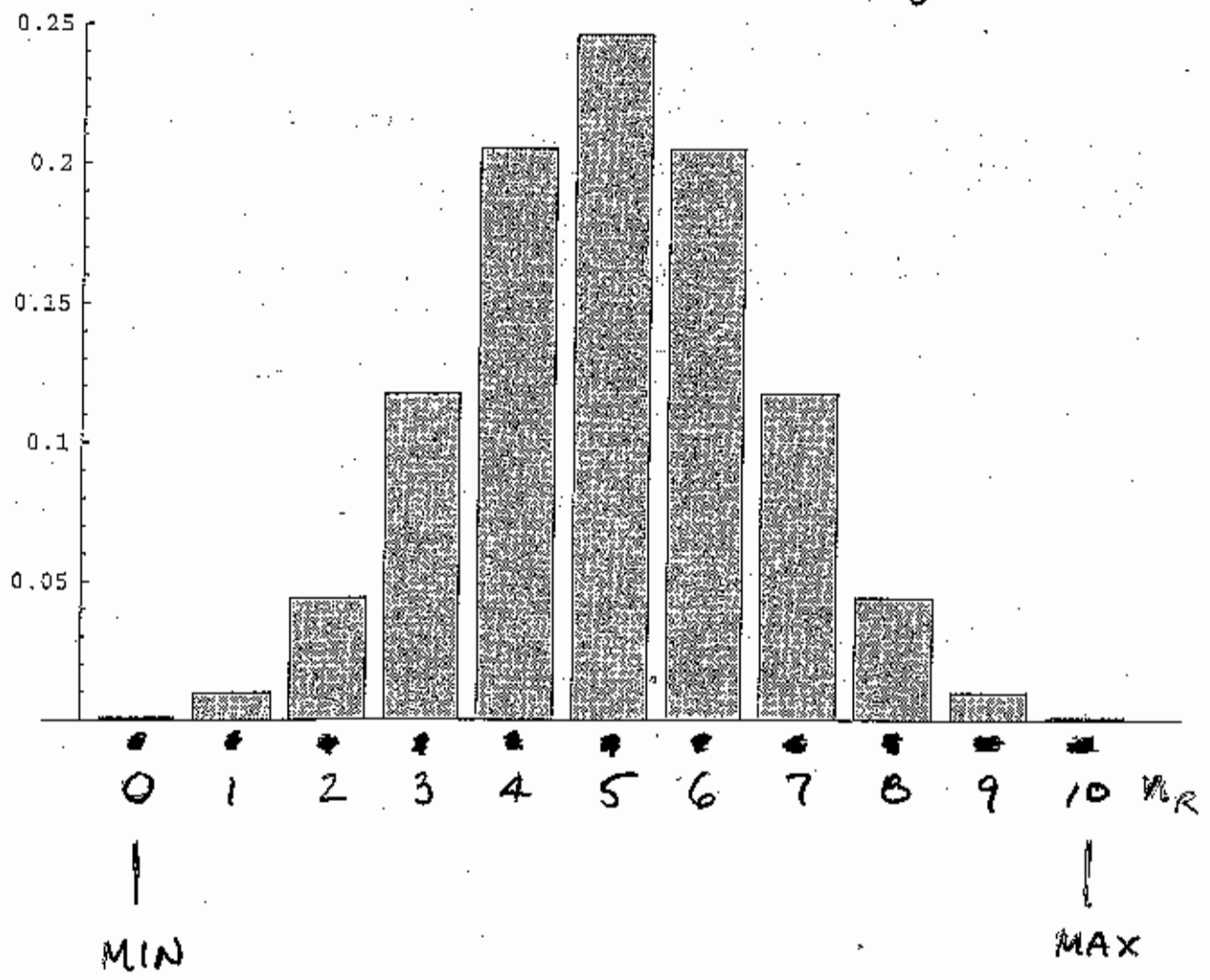
$$\frac{\Delta^*x}{\bar{x}} \sim \frac{1}{N^{1/2}}$$

(fn sits at 0 for most x , jumps quickly near peak)
cf plots

$W(n_R)$

$N = 10$

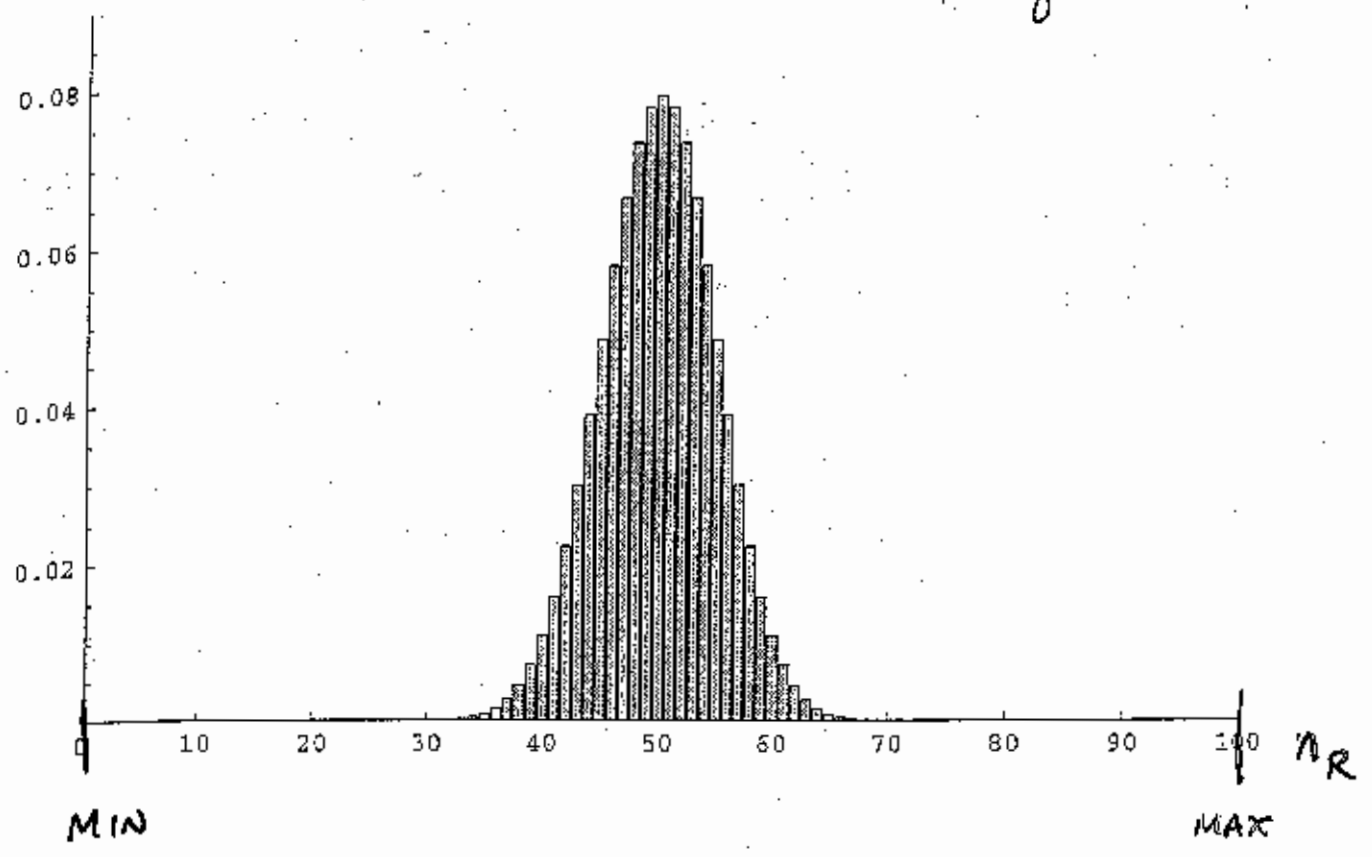
$p = q = \frac{1}{2}$



$W(n_R)$

$N = 100$

$p = q = \frac{1}{2}$



FN IS EXPONENTIAL:

POWERS ARE EXPONENTIAL: $p^{nr} = e^{nr \ln p}$

C.4

LARGE FACTORIALS \approx EXPONENTIALS

(try 100! on calculator)

STIRLING APPROX (cf appx)

$$n! \approx e^{n \ln n - n + \frac{1}{2} \ln(2\pi n) + \dots}$$

$$= (2\pi n)^{\frac{1}{2}} e^{n(\ln n - 1)}$$

ex $100! \sim e^{364}$

terms which
vanish as $n \rightarrow \infty$

(uses method we'll
talk about now)

"leading term in
asymptotic expansion"

PLAN: TO FIND SIMPLE APPROX

(1) GOOD NEWS:

$P(x)$ ONLY ≈ 0 NEAR PEAK \tilde{x}

\Rightarrow ONLY NEED GOOD APPROX NEAR MAX

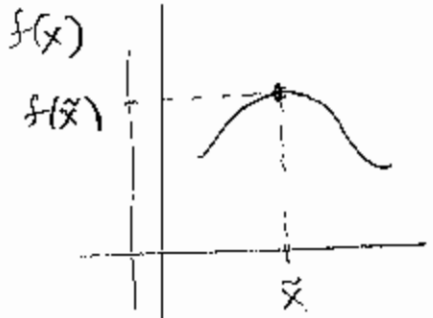
\Rightarrow TAYLOR SERIES

(2) BAD NEWS:

$P(x)$ DROPS EXPONENTIALLY

\Rightarrow BUILD INTO SERIES

SIMPLE APPROX. NEAR PEAK:



TAYLOR:

$$f(x) \sim f(\bar{x}) + \underbrace{f'(\bar{x})}_{=0} (x - \bar{x}) + \underbrace{\frac{1}{2} f''(\bar{x})}_{\leq 0} (x - \bar{x})^2 + \dots$$

(EXTREMUM) (CURVES DN)



DEFINE $\eta \equiv x - \bar{x}$

(SO EXPAND AROUND $\eta = 0$)

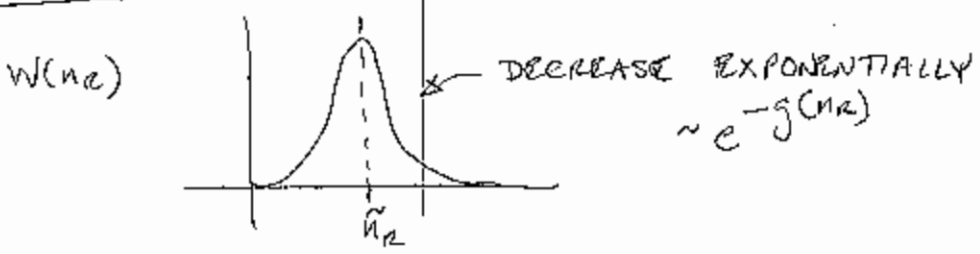
$$f(x) \equiv f(\bar{x} + \eta) \approx f(\bar{x}) + \frac{1}{2} f''(\bar{x}) \eta^2 + \dots$$

$$= f(\bar{x}) - \frac{1}{2} |f''(\bar{x})| \eta^2 + \dots$$

ACCURATE FOR η SMALL

TEST: NEW TERMS GET SMALLER

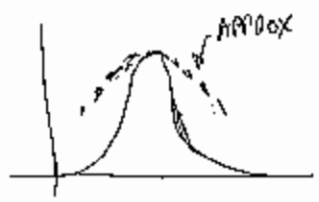
OUR CASE



TAYLOR SERIES: NO POLYNOMIAL CAN KEEP UP

WITH EXPONENTIAL FOR LONG \Rightarrow ONLY GOOD FOR SMALL RANGE IN $\eta \equiv nr - \bar{n}_r$

\Rightarrow GET



BE SMART:

KNOW $W(n_r) \sim e^{-g(n_r)}$

APPROX g , NOT W (builds in exp behavior)

ie $W(n_r) = e^{\frac{\ln W(n_r)}{\quad}}$
 \rightarrow MUCH SMOOTHER FN
 \rightarrow EASIER TO APPROX

TAYLOR SERIES:

$$(w \ \eta \equiv n_r - \tilde{n}_r)$$

$$\ln W(n_r) \approx \ln W(\tilde{n}_r) + B_1 \eta + \frac{1}{2} B_2 \eta^2 + \frac{1}{3!} B_3 \eta^3 + \dots$$

$$B_k = \frac{d^k (\ln W(\tilde{n}_r))}{d n_r^k}$$

KNOW

$$B_1 = 0$$

$$B_2 < 0$$

THEN

$$W(n_r) \approx e^{\left[\ln W(\tilde{n}_r) - \frac{1}{2} |B_2| \eta^2 + \dots \right]}$$

$$\approx W(\tilde{n}_r) e^{-\frac{1}{2} |B_2| \eta^2}$$

"GAUSSIAN DISTR."

 \rightarrow WILL CLEAN UP LATER \Rightarrow VERY GENERAL:

^{ONLY}
 \rightarrow ASSUMED EXP. FALL OFF

\rightarrow VERY COMMON WHEN ADD LARGE $\neq N$
 OF ~~STATIST.~~ ^{RANDOM} VARIABLES (HERE N STEPS OF $s_i = \pm 1$
 WITH PROB p, q)

("CENTRAL LIMIT THM" \rightarrow
 ALMOST ANY PROB FOR INDIV. s_i GIVES THIS FORM)

RANGE OF VALIDITY:

- GOOD APPROX FOR $\ln W$ IF

$$\left| \frac{1}{3} B_3 \eta^3 \right| \ll \left| \frac{1}{2} B_2 \eta^2 \right|$$

it fails for large enough η

- WILL SEE: WHEN " \sim "
 W NEGLIGIBLE

\Rightarrow APPROX GOOD FOR ALL η

ex (REIF):

$$f(y) = \frac{1}{(1+y)^N} = e^{-N \ln(1+y)}$$

{ powers,
exponentials
~ same

→ CHANGES SHARPLY IF N LARGE } see picture
 → $\ln f(y)$ MUCH SMOOTHER }

APPROX NEAR $y=0$

DIRECTLY: $f(y) \sim 1 - Ny + \frac{1}{2}N(N+1)y^2 + \dots$

N large → NOT SENSIBLE UNLESS $|Ny| \lesssim 1$
 OR $|y| \lesssim \frac{1}{N}$

LOG: $\ln f(y) \sim -N(y - \frac{1}{2}y^2 + \dots)$

OK IF $y^2 \lesssim y$ OR $y \lesssim 1$

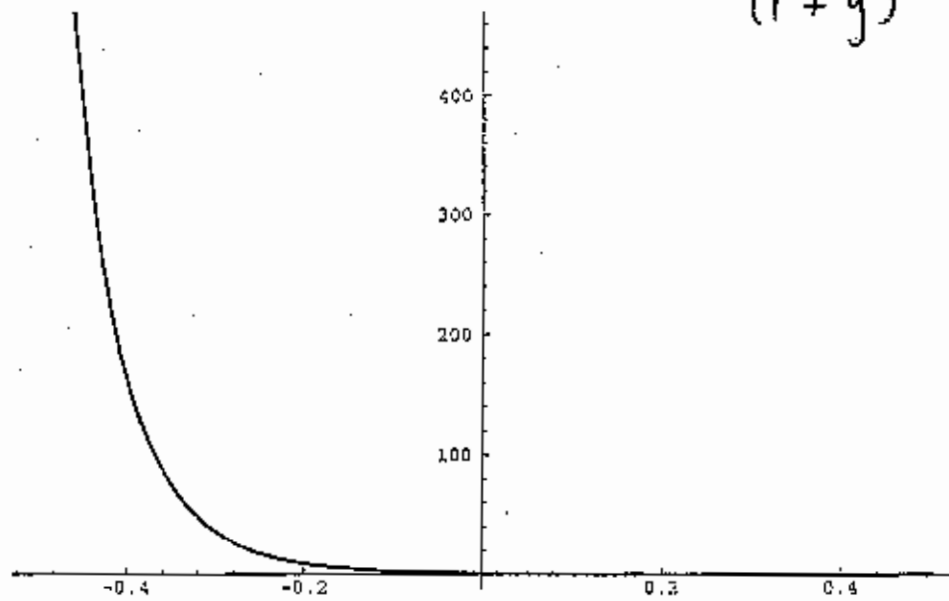
$f(y) \sim e^{-N(y - \frac{1}{2}y^2 + \dots)}$ (indep of N)

⇒ picture:

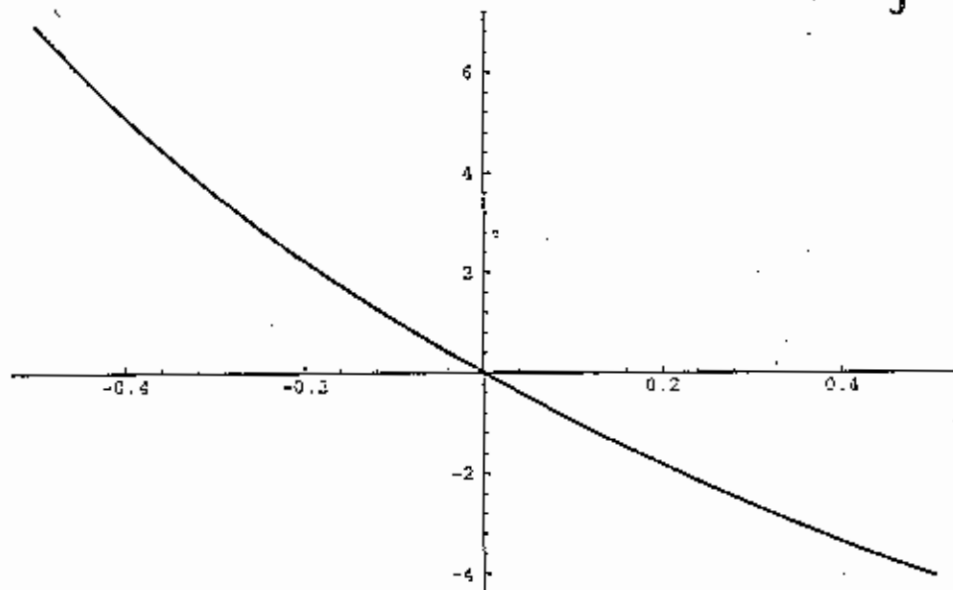
$N = 10$

1st pict dies at $y \sim \pm \frac{1}{10} = \pm 0.1$

$$\frac{1}{(1+y)^{10}}$$



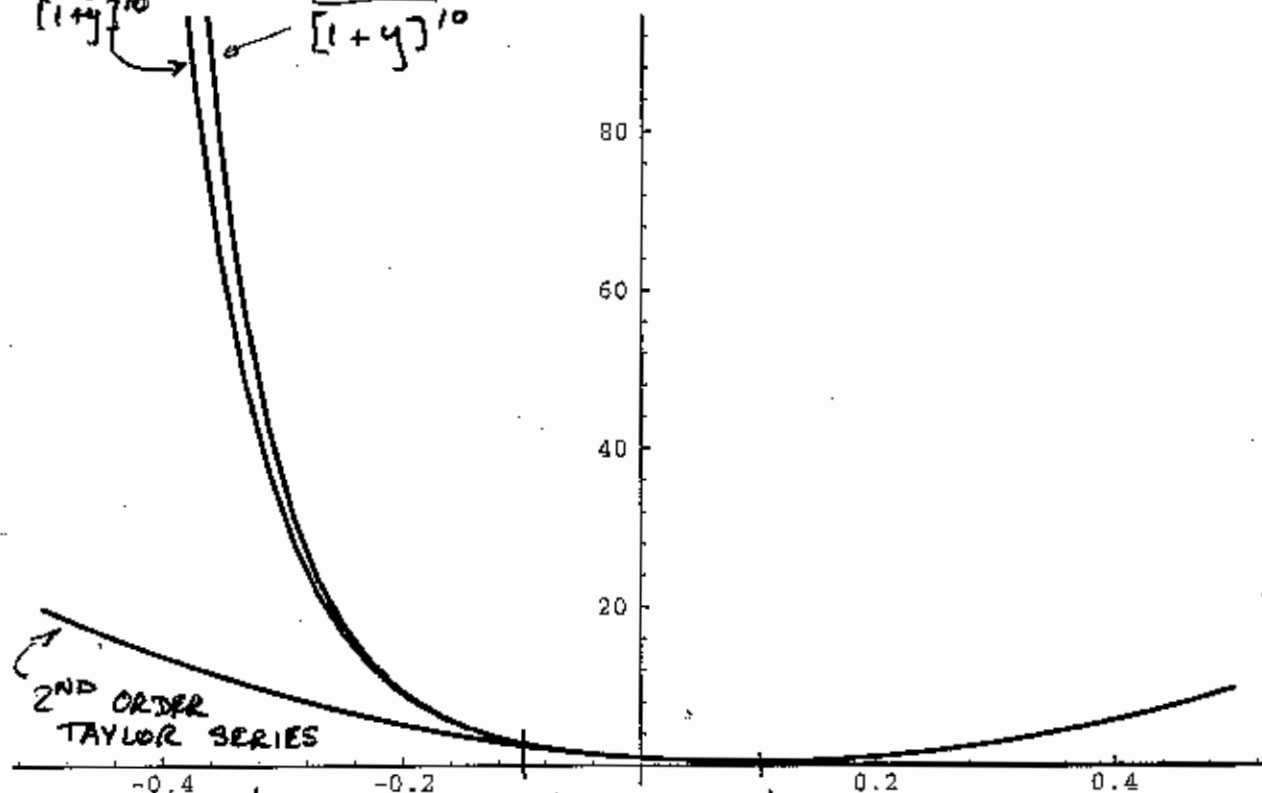
$$\ln \left[\frac{1}{(1+y)^{10}} \right]$$



2ND ORDER
MATCH TO LOG
OF

$$\frac{1}{[1+y]^{10}}$$

$$\frac{1}{[1+y]^{10}}$$



2ND ORDER
TAYLOR SERIES

FOR
 $\frac{1}{[1+y]^{10}}$

$$-0.1 = \frac{1}{N}$$

APPROX $\ln W(n_R)$: (cf REIF FOR DETAILS
(or try yourself))

PROBLEM: NEED DERIVS

ONLY DEF'D AT INTEGER n_R 'S

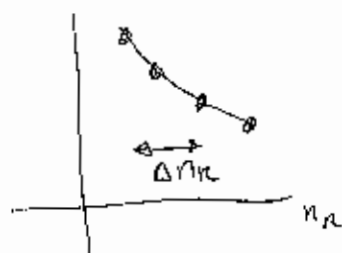
BEST APPROX:

$$\frac{d \ln W(n_R)}{d n_R} \approx \frac{\Delta \ln W(n_R)}{\Delta n_R}$$

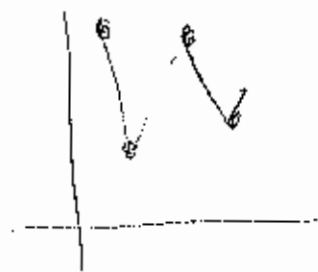
$$\text{SMALLEST } \Delta n_R = 1$$

$$= \ln W(n_R+1) - \ln W(n_R)$$

OK IF $\ln W(n_R)$ SMOOTH:



VS



(see in plot that W becomes smoother as N incr.)

(a) MAX: $\frac{d \ln W}{d n_R} = 0$ (*)

SOLN: $\tilde{n}_R = N_p$ (note: same as \bar{n}_R)

(*) NEED THINGS LIKE

$$\frac{\Delta \ln n!}{\Delta n} \sim \frac{\ln(n+1)! - \ln n!}{1} = \ln \left[\frac{(n+1)!}{n!} \right] = \ln(n+1) \sim \ln n$$

$$\left\{ \text{or } \ln(n+1) + \ln n + \ln(n-1) + \dots - \ln(n) - \ln(n-1) + \dots \right\}$$

$$\Delta n \gg 1 \Rightarrow \frac{d \ln n!}{d n} \sim \ln n \quad \text{ETC}$$

APPROX $\ln W(n_R)$: (cf REIF FOR DETAILS or better, try yourself)

FIND \tilde{n}_R AT MAX:

(a) $\frac{d \ln W(n_R)}{d n_R} = 0 \approx \ln W(n_{R+1}) - \ln W(n_R)$

SOLN $\tilde{n}_R = N_p$

(b) EXPAND AROUND \tilde{n}_R ($\eta = n_R - \tilde{n}_R$)

COEFFICIENTS:

$B_1 = 0$
 $B_2 = -\frac{1}{N_p g}$

AS EXPECTED, < 0

$B_3 = \frac{g^2 - p^2}{N^2 p g^2}$

ETC

EACH NEW TERM $B_k \eta^k$ HAS ADDITIONAL FACTOR

$\frac{\eta}{N_p g}$

TIMES PREVIOUS;

SERIES

OK IF $\frac{\eta}{N_p g} \ll 1$

skip; done later

skip (N BIG; p, g NOT TOO SMALL; CF PROB 1.9) (and when $\eta \sim N_p g$, $W \sim 0$)

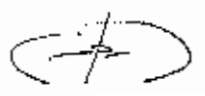
(c) $W(\tilde{n}_R) \cong \frac{1}{(2\pi N_p g)^{1/2}}$

(STIRLING FORMULA) (1st lecture, could also use more cond) APPX

FINALLY $W(n_R) = W(\tilde{n}_R) e^{-\frac{1}{2} |B_2| \eta^2}$

$= \frac{1}{(2\pi N_p g)^{1/2}} e^{-\frac{(n_R - N_p)^2}{2 N_p g}} + \left[\text{CORRECTIONS WHICH VANISH AS } N \text{ LARGE} \right]$

(using (a) & (b) above)



{ formula:

$W(\tilde{n}_k)$ using Stirling's formula:

$$W(n_k) = \frac{N!}{(N-n_k)! n_k!} p^{n_k} q^{N-n_k}$$

$$\tilde{n}_k = Np$$

$$W(\tilde{n}_k) = \frac{N!}{(N-Np)! (Np)!} p^{Np} q^{N-Np}$$

$$= \frac{N!}{(Ng)! (Np)!} p^{Np} q^{Ng}$$

$$\ln W(\tilde{n}_k) = \ln N! - \ln(Ng)! - \ln(Np)! + Np \ln p + Ng \ln q$$

Stirling:

$$\ln n! \sim n \ln n - n + \frac{1}{2} \ln(2\pi n) + \dots$$

"large"

$$\ln W(\tilde{n}_k) \sim N \ln N - N + \frac{1}{2} \ln(2\pi N)$$

$$- Ng \ln Ng + Ng - \frac{1}{2} \ln(2\pi Ng)$$

$$- Np \ln Np + Np - \frac{1}{2} \ln(2\pi Np)$$

$$+ Np \ln p + Ng \ln q$$

$$\text{from } p+q=1 \Rightarrow -Ng \ln Ng - Np \ln Np = \frac{-Ng \ln N - Np \ln N - Ng \ln q - Np \ln p}{-N \ln N}$$

everything cancels but

$$\frac{1}{2} [\ln(2\pi N) - \ln(2\pi Ng) - \ln(2\pi Np)] = \frac{1}{2} \ln \frac{2\pi N}{(2\pi Ng)(2\pi Np)}$$

$$= \frac{1}{2} \ln \frac{1}{2\pi Npq}$$

$$\Rightarrow W(\tilde{n}_k) = \left(\frac{1}{2\pi Npq} \right)^{\frac{1}{2}}$$

}

THEN $W(n_R) \approx W(\tilde{n}_R) e^{-\frac{1}{2}|B_2|\eta^2}$

USE $W(\tilde{n}_R) \equiv \frac{1}{(2\pi N p q)^{1/2}}$ (STIRLING'S FORMULA)

$$|B_2| = \frac{1}{N p q}$$

AND $\eta \equiv n_R - \tilde{n}_R = n_R - N p$

$$W(n_R) = \frac{1}{(2\pi N p q)^{1/2}} e^{-\frac{(n_R - N p)^2}{2 N p q}} + \text{CORRECTIONS} \sim \frac{1}{N}$$

CONVERT TO $P(x) = \frac{P(m)}{2l}$

USE $n_R = \frac{1}{2}(N+m)$, $x = ml$

ALSO $\mu \equiv (p-q)Nl$ (just defns, but we'll see $\mu = \bar{x}$, $\sigma^2 = \overline{(Ax)^2}$)
 $\sigma^2 \equiv 4 N p q l^2$ (put on side)

THEN $P(x) = \frac{1}{(2\pi\sigma^2)^{1/2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

GAUSSIAN DISTRIBUTION

(discuss - peaked at $x \sim \mu$, falls off for $|x-\mu| > \sigma$)

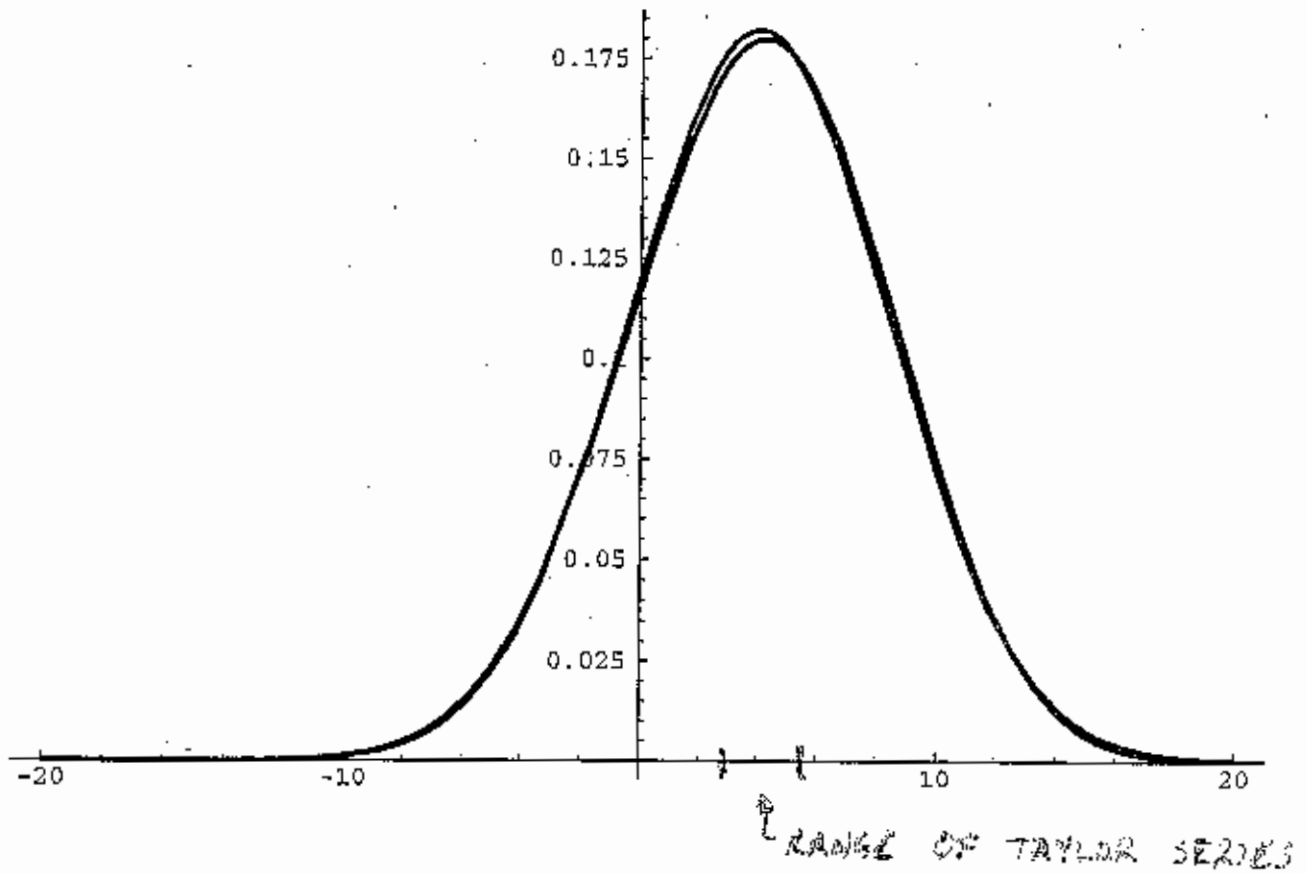
ALMOST ALWAYS GET THIS WHEN ADD LARGE # RAND. VARS. (CENTRAL LIMIT THM - cf # SECTIONS FOR PROOF)

μ & σ HAVE SIMPLE INTERPRETATION

skip
incl plot

$N=20$ $p=.6$

GAUSSIAN VS BINOMIAL
DISTRIBUTIONS
(PLOTTED AS
CONTINUOUS FNS)



CHECK NORMALIZATION:

$$\int_{x_{\min}}^{x_{\max}} P(x) dx$$

↑ LIMITS: $\pm N\sigma$

$$P(x) \approx 0$$

REPLACE w/ $\pm \infty$ (only affects by terms which vanish
edg. w/ N)

NEED

$$\int_{-\infty}^{\infty} P(x) dx = \frac{1}{(2\pi\sigma^2)^{1/2}} \int_{-\infty}^{\infty} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$y = x - \mu$$

$$\int_{-\infty}^{\infty} e^{-y^2/2\sigma^2} dy$$

2 TRICKS: DO ALL GAUSS. INTS (one reason they're so popular -
ints are easy)

CONSIDER $I(\alpha) = \int_{-\infty}^{\infty} e^{-\alpha z^2} dz$

COULD DO IF HAD $z dz$

$$\begin{aligned} I^2(\alpha) &= \int_{-\infty}^{\infty} e^{-\alpha z_1^2} dz_1 \int_{-\infty}^{\infty} e^{-\alpha z_2^2} dz_2 \\ &= \int_{-\infty}^{\infty} dz_1 dz_2 e^{-\alpha(z_1^2 + z_2^2)} \end{aligned}$$

↓ (ONLY DISCUSS IF NEEDED
IMMEDIATELY FOR PROB 15)

1.24.1

INSERT: (put this here to summarize for HW; can drop next time)

NORM:

$$x_{\text{MAX}} = +N\ell$$

$$x_{\text{MIN}} = -N\ell$$

$$P(x) dx$$

~ 0 long before MIN, MAX

$$\approx \int_{-\infty}^{\infty} P(x) dx = \int_{-\infty}^{\infty} \frac{1}{(\sqrt{2\pi}\sigma)^2} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

CAN SEE = 1 (later?)

PROB. $x < y$?

y

$$\int_{-\infty}^y P(x) dx$$

\Rightarrow ERROR FN, APPX A.5

PROBLEM 15: FOR WHAT y IS PROB. $> C$?

OF TABLES, MATHEMATICA (ask teacher)

WILL SEE:

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

$$\bar{x} = \int P(x) \cdot x dx = \mu$$

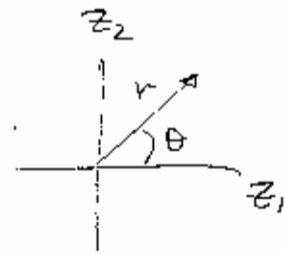
$$\overline{(\Delta x)^2} = \int P(x) (x - \bar{x})^2 dx = \sigma^2$$

} PARAMS
IN $P(x)$

↓ (later)

POLAR COORDS

$$\text{skip } \left\{ \begin{array}{l} r = (z_1^2 + z_2^2)^{1/2} \\ \tan \theta = z_2 / z_1 \end{array} \right.$$



$$dz_1 dz_2 \rightarrow r dr d\theta \quad z_1^2 + z_2^2 = r^2$$

$$I^2(\alpha) = \int_0^{2\pi} d\theta \int_0^{\infty} r dr e^{-\alpha r^2}$$

$\frac{1}{2\alpha} \quad u = r^2 \quad du = 2r dr$

$$I^2(\alpha) = \frac{\pi}{\alpha}$$

$$I(\alpha) = \left(\frac{\pi}{\alpha}\right)^{1/2}$$

CAN USE THIS TO COMPUTE RELATED INTS:

$$\begin{aligned} \frac{\partial}{\partial \alpha} I(\alpha) &= \frac{\partial}{\partial \alpha} \int_{-\infty}^{\infty} e^{-\alpha z^2} dz = \frac{\partial}{\partial \alpha} \left(\frac{\pi}{\alpha}\right)^{1/2} \\ &= - \int_{-\infty}^{\infty} e^{-\alpha z^2} z^2 dz = -\frac{1}{2} \left(\frac{\pi}{\alpha^3}\right)^{1/2} \end{aligned}$$

$$\frac{\partial^n}{\partial \alpha^n} I(\alpha) = (-1)^n \int_{-\infty}^{\infty} e^{-\alpha z^2} z^{2n} dz = \frac{\partial^n}{\partial \alpha^n} \left(\frac{\pi}{\alpha}\right)^{1/2}$$

ALSO

$$\int_{-\infty}^{\infty} e^{-\alpha z^2} z^{2n+1} dz = 0$$

(z ODD
 $\Rightarrow \pm z$ GIVE OPP
 CONTRIB)

USE SIMILAR TRICK FOR:

1.26

$$\int_0^{\infty} e^{-\alpha z^2} z^{2n+1} dz$$

(EVEN ARE JUST $\frac{1}{2}$ OF $\int_{-\infty}^{\infty}$ INTS)

START WITH $\int_0^{\infty} e^{-\alpha z^2} z dz = \frac{1}{2\alpha}$

THEN $\boxed{(-1)^n \int_0^{\infty} e^{-\alpha z^2} z^{2n+1} dz = \frac{\partial^n}{\partial \alpha^n} \frac{1}{2\alpha}}$ n=0,1,2,...

IN PARTICULAR:

$$\int_{-\infty}^{\infty} P(x) dx = \frac{1}{(2\pi\sigma^2)^{1/2}} \int_{-\infty}^{\infty} e^{-y^2/2\sigma^2} dy$$

here $\alpha = \frac{1}{2\sigma^2}$ $\left(\frac{\pi}{\alpha}\right)^{\frac{1}{2}} = (2\pi\sigma^2)^{1/2}$

= 1 ✓

ALSO $\bar{X} = \int_{-\infty}^{\infty} P(x) \cdot x dx$
 $\begin{cases} \frac{1}{(2\pi\sigma^2)^{1/2}} \int_{-\infty}^{\infty} e^{-(x-\mu)^2/2\sigma^2} x dx \\ \text{(a) } y = x - \mu \\ \text{(b) } \alpha = \frac{1}{2\sigma^2} \end{cases}$
 $\left. \begin{matrix} (y + \mu) \\ \text{control} \\ \text{norm.} \end{matrix} \right\}$

$\Rightarrow \boxed{\bar{X} = \mu}$ (~ AS EXPECTED)

$$\overline{(\Delta X)^2} = \overline{(X - \bar{X})^2} = \int_{-\infty}^{\infty} \underbrace{(x-\mu)^2}_{= y^2} \underbrace{P(x)}_{\frac{1}{(2\pi\sigma^2)^{1/2}} e^{-y^2/2\sigma^2}} dx$$

$\boxed{\overline{(\Delta X)^2} = \sigma^2}$ $\frac{1}{2} \left(\frac{\pi}{\alpha^3}\right)^{\frac{1}{2}} \alpha = \frac{1}{2\sigma^2}$

∴ (1) CAN READ OFF \bar{X} & $\overline{(\Delta X)^2}$ FROM GAUSS DISTN: (ie μ AND σ^2)

(2) HERE $\bar{X} = (p-q)Nl = \bar{m}l$

$\overline{(\Delta X)^2} = 4Npq l^2 = \overline{(\Delta m)^2} l^2$

∴ (AS IT MUST) (I suppose up to $\sim \frac{1}{N}$)

$\left. \begin{matrix} \text{consistent} \\ \text{w/ results} \\ \text{from disc. dist} \end{matrix} \right\}$

ONE LAST INTEGRAL: (NOT ALL FUN & GAMES)

ERROR FN: (cf prob 15)

def FOR $P(x) = \frac{1}{(2\pi\sigma^2)^{\frac{1}{2}}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$

ex FIND PROB $x < y$: (OR $y_1 < x < y_2$)

$$\int_{-\infty}^y P(x) dx$$

RELATED TO

$$\text{erf}(y) \equiv \frac{2}{\sqrt{\pi}} \int_0^y e^{-x^2} dx \quad \left(\int_{-\infty}^0 \text{ part is easy} \right)$$

cf APPX A.5 (DISCUSSION & APPROX. FORMULA)

⇒ NO CLOSED FORM; DO NUMERICALLY OR
LOOK UP IN TABLES, MATHEMATICA ($\text{Erf}[y]$), C



VALIDITY OF GAUSSIAN APPROX:

$$W(n_R) \approx e^{\left[\ln W(\tilde{n}_R) + \frac{1}{2} B_2 \eta^2 + \frac{1}{3!} B_3 \eta^3 + \dots \right]}$$

skip $\left\{ \begin{array}{l} \tilde{n}_R \text{ IS PT WHERE } W \text{ IS MAX} \\ B_k \equiv \frac{d^k (\ln W)}{dn_R^k} \Big|_{n_R = \tilde{n}_R} \end{array} \right. \quad \eta \equiv n_R - \tilde{n}_R$

$B_1 = 0$ (MAX)
 $B_2 = -\frac{1}{Npg}$ ($< 0 \rightarrow$ MAX)

EACH NEW TERM:

EXTRA $\frac{\eta}{Npg}$ (ex $B_3 = \frac{p^2 - p^2}{N^2 p^2 g^2}$)

(a) SERIES OK WHILE $\frac{\eta}{Npg} \ll 1$ OR $\eta \ll Npg$

skip $\Rightarrow W(n_R) = W(\tilde{n}_R) e^{-\frac{1}{2} |B_2| \eta^2}$ GOOD APPROX

ie: N large
 P, g NOT TOO SMALL
 (cf para 9
 for diff approx
 when p or g
 small)

(b) FAILS WHEN $\eta \approx Npg$

HOW BIG IS W?

$W \approx W(\tilde{n}_R) e^{-\frac{1}{2} \frac{1}{Npg} (Npg)^2}$
 $e^{-\frac{1}{2} Npg}$

IF $Npg \gg 1$, $W \sim 0$, APPROX GOOD FOR ALL η
 (ie it's already zero when fails)

USEFUL GENERAL RESULTS (SECT 1.9)

(more general than ran. walk, Brown approx)

CONSIDER SUM OF RAN. #s S_i (N NOT NECESS LARGE)

$$X = S_1 + S_2 + S_3 \dots + S_N = \sum_{i=1}^N S_i$$

EX FINAL POSITION AFTER N STEPS

$$S_1 = \begin{matrix} 1^{\text{st}} \text{ STEP} = +l & \text{w/ PROB } p \\ & -l & \text{" " } q \end{matrix}$$

{ INDEX ON S_i HERE IS WHICH STEP, NOT WHAT VALUES S CAN TAKE

($W \equiv$ doublet, not omega)

HERE: ASSUME ONLY SAME DISTR. $W(S)$ FOR EACH S_i , \dagger INDEP.

- DISCRETE OR CONTINUOUS

WHAT IS \bar{X} ? HOW DOES IT FLUCTUATE ?

{ because S_i 's random X is also random, means it has some distr. $P(x)$

(A) AVE OF SUM = SUM OF AVES (have seen already)

AGAIN:

$$\bar{X} = \frac{1}{M} \sum_{m=1}^M X_m \quad \underline{M \rightarrow \infty}$$

$$= \frac{1}{M} \left\{ (S_1 + S_2 + \dots + S_N) + \right. \quad \left. \begin{matrix} \{ m=1 \\ \\ \\ \} \end{matrix} \right.$$

$$+ (S_1' + S_2' + \dots + S_N') + \quad \left. \begin{matrix} \{ m=2 \\ \\ \\ \} \end{matrix} \right.$$

$$+ (S_1^{(M)} + S_2^{(M)} + \dots + S_N^{(M)}) \quad \left. \begin{matrix} \{ m=M \} \end{matrix} \right\}$$

$$= \bar{S}_1 + \bar{S}_2 + \dots + \bar{S}_N$$

{ here: N is just a sum - could be 3, for ex, but M is to average $\Rightarrow \infty$ }

ALSO, EACH STEP HAS SAME PROB DIST: $\bar{s}_1 = \bar{s}_2 = \dots$

$$\bar{x} = \frac{\sum_{i=1}^N \bar{s}_i}{N} = N \bar{s}$$

{ emphasizing diff between $\sum_{i=1}^N$ and sum used in ave

$$w/ \bar{s} = \int ds w(s) \cdot s$$

{ if $w(s)$ continuous

else sum: ex ran walk
 $\bar{s} = p \cdot l + q \cdot (-l)$
 $= (p - q)l$

(b) SPREAD ?

$$\sum_{i=1}^N s_i - N \bar{s}$$

$$\Delta X = X - \bar{X} = \sum_{i=1}^N (s_i - \bar{s}) = \sum_{i=1}^N \Delta s_i$$

$$\overline{(\Delta X)^2} = \overline{\left(\sum_{i=1}^N \Delta s_i \right) \left(\sum_{j=1}^N \Delta s_j \right)}$$

(need diff i, j to keep two Δx 's straight; write out if necessary)

SEPARATE OUT DIAG, CROSS TERMS

$$= \overline{\sum_{i=j=1}^N (\Delta s_i)^2} + \overline{\sum_{i \neq j} (\Delta s_i)(\Delta s_j)}$$

ave of sum = sum of aves

$$\begin{array}{ccc} \parallel & & \parallel \\ \sum (\Delta s_i)^2 & & \sum (\Delta s_i)(\Delta s_j) \\ \parallel & & \parallel \\ \text{each} & & \\ \text{same} & & \\ N (\Delta s)^2 & & \end{array}$$

INSIDE 2ND SUM: IS AVE OF PROD = PROD OF AVE ?

\Rightarrow ONLY IF 2 TERMS STATIST. INDEP:

IF a, b ^{INDEP} RANDOM. #S w/ DISTR. $P(a), Q(b)$

PROB $a \ \& \ b$: $P(a) \cdot Q(b)$

(NOT INDEP? SOME OTHER GEN'L DISTR. $R(a, b)$
GIVES PROB OF BOTH; TAKES AFFECT OF a ON b INTO ACCT)

THEN:

$$\begin{aligned}\overline{ab} &= \sum_{k,l} P(a_k) Q(b_l) (a_k b_l) \\ &\quad \underbrace{\hspace{10em}}_{\text{all possible values for } a \cdot b} \\ &= \left(\sum_k P(a_k) a_k \right) \left(\sum_l Q(b_l) b_l \right) \\ &= \overline{a} \cdot \overline{b}\end{aligned}$$

OUR PROBLEM: EACH TERM ($i \neq j$)

$$\overline{(\Delta S_i)(\Delta S_j)} = \underbrace{\overline{(\Delta S_i)}}_{=0} \cdot \underbrace{\overline{(\Delta S_j)}}_{=0} = 0$$

(wouldn't work for $i=j \Rightarrow$ certainly not indep)

FINALLY

$$\overline{(\Delta X)^2} = N \overline{(\Delta S)^2} \quad \text{OR} \quad \boxed{\Delta^* X = \sqrt{N} \Delta^* S}$$

MAKES SENSE: MORE TERMS (N SUM, MORE TOTAL MOVES AROUND)

BUT: BIGGER N , SMALLER RELATIVE CHG:

$$\boxed{\frac{\Delta^* X}{\bar{X}} = \frac{1}{\sqrt{N}} \frac{\Delta^* S}{\bar{S}}}$$

WHY IS THIS USEFUL?

(0) CAN TELL ME, STD DEV FOR SUM ONLY KNOWING μ, σ FOR PIECES; TELLS A LOT ABOUT DISTR.

(1) N LARGE: (CAN SAY MORE)

SPTS S HAS ARBITRARY DISTR $w(s)$ & N LARGE

(i.e. RANDOM WALK)

step $\left\{ \begin{array}{l} \rightarrow \text{COMPUTE } \bar{s}, \Delta^* s = \sqrt{(\Delta s)^2} \end{array} \right.$

ALSO $X \equiv s_1 + \dots + s_N, \quad \underline{\underline{N \text{ LARGE}}}$

THEN: KNOW DISTRIB. $P(x)$ FOR x COMPLETELY

How?

(a) $\bar{x} = N \bar{s} \equiv \mu$

(b) $\left\{ \begin{array}{l} \overline{(\Delta x)^2} = \overline{N(\Delta s)^2} \\ \text{OR} \end{array} \right. \Delta^* x = \sqrt{N} \Delta^* s = \sigma$

(c) $N \text{ LARGE} \Rightarrow P(x) \text{ GAUSSIAN}$ } of optimal sections 10⁸!!

$\Rightarrow P(x) = \frac{1}{(2\pi\sigma^2)^{\frac{1}{2}}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$ "central limit theorem"

\Rightarrow CAN COMPUTE ANYTHING ABOUT x

\Rightarrow TRUE FOR ANY $w(s)$ (w/ some rare exceptions)

(2) STATISTICAL ERRORS:

ACCURACY OF MEAS. IMPROVES AS $\frac{1}{\sqrt{N}}$

IF REPEAT N TIMES: (if errors are statistical)

WHY:

MEAS. QTY s N TIMES

BEST VALUE: $S_{\text{BEST}} = \frac{1}{N} \sum_{i=1}^N s_i \equiv \frac{1}{N} X$

($i = \#$ OF MEASMT.)

WHAT SHOULD QUOTE FOR ERROR?

$$\Rightarrow \frac{1}{\sqrt{N}} \Delta^* S$$

WHY?

HOW MUCH DOES S_{BEST} FLUCT?

$$\Delta^* S_{\text{BEST}} = \frac{1}{N} \Delta^* X = \frac{1}{N} (\sqrt{N} \Delta^* S) = \frac{1}{\sqrt{N}} \Delta^* S$$

ie: gives idea of what std dev. would be
if repeated (N meas, then ave) large
times.

DISTRIBUTION FOR S_{BEST} ?

IF LARGE N : \Rightarrow GAUSSIAN

WORK 10 TIMES HARDER \Rightarrow DO ~ 3 TIMES BETTER
(it more accurate)

