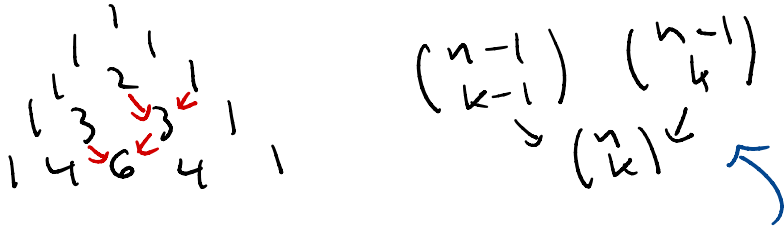


Most important pattern in Pascal's  $\Delta$  is **Pascal's identity**:

Thm  $\binom{n}{k} = \binom{n-1}{k} + \binom{n-1}{k-1}$

e.g.



Note: Let's you easily **fill in** Pascal's  $\Delta$ !

Pf: Let's define a **bijection**

$$f: \{ \text{k-subsets of } [n] \} \rightarrow \{ \text{k-subsets of } [n-1] \} \cup \{ \text{k-1-subsets of } [n-1] \}$$

$$\text{by } f(A) = \begin{cases} A & \text{if } n \notin A \text{ (a k-subset of } [n-1]) \\ A \cup \{n\} & \text{if } n \in A \text{ (a k-1-subset of } [n-1]) \end{cases}$$

This exactly corresponds to Pascal's identity.  $\square$

Rmk: We have a similar identity

$$S(n, k) = S(n-1, k-1) + k \cdot S(n-1, k)$$

for Stirling #'s of 2<sup>nd</sup> kind, w/ a very similar bijective proof. //

# Math 4707: More Pascal's triangle and probability

2/3  
ch's 3+5  
of LPV

Reminder: HW #1 is due **today**!

Last class we introduced **Pascal's triangle** of  $\binom{n}{k}$ :

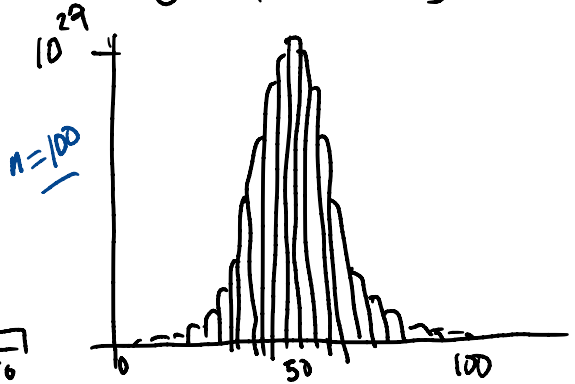
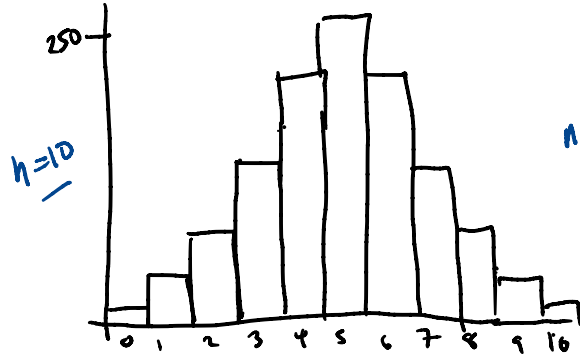
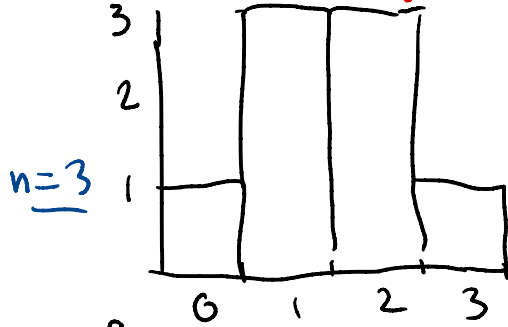
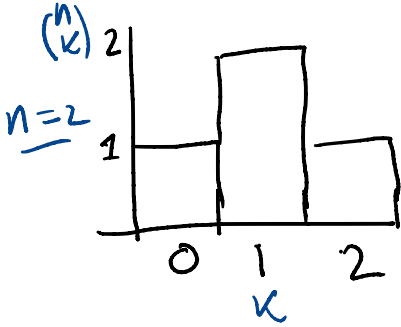
$$\begin{array}{cccccc} & & & 1 & & & \\ & & & & 1 & & \\ & & 1 & & 2 & & 1 \\ & 1 & & 3 & & 3 & & 1 \\ 1 & & 4 & & 6 & & 4 & & 1 \\ & & & & \vdots & & & & \end{array}$$

and discussed various **patterns** in it, like **symmetry**, the **sum/alternating sum** of a row, and, most important, **Pascal's identity**  $\binom{a}{n} + \binom{b}{n} = \binom{a+b}{n}$ . We will discuss more patterns like this on the **worksheet** for today.

But in today's lecture, instead we're going to talk about the **large scale** behavior of Pascal's  $\Delta$ , and its connections to **basic probability**. The material for today is mostly **"cultural"**, i.e., I will not assess you on it. However, it is still very interesting + important.

Q: What does the  $n^{\text{th}}$  row of Pascal's  $\Delta$  roughly "look like," for big  $n$ ?

To answer this, helpful to draw a **histogram**:



What do we see in these pictures?

- **symmetry**  $\binom{n}{k} = \binom{n}{n-k}$  ✓
  - numbers get **bigger** towards the **middle**
- Indeed,

$$\binom{n}{k} < \binom{n}{k+1} \iff \frac{n!}{k!(n-k)!} < \frac{n!}{(k+1)!(n-k-1)!}$$

$$\Leftrightarrow 1 < \frac{n-k}{k+1}$$

$$\Leftrightarrow k < \frac{n-1}{2},$$

So for first half of  $k$ , have  $\binom{n}{k} < \binom{n}{k+1}$  ✓

• middle number is **pretty big**

Recall sum of  $\binom{n}{k} = 2^n$ , and there are  $n+1$   $k$ 's

$$\Rightarrow \text{average over } k \binom{n}{k} = \frac{1}{n+1} 2^n$$

$$\Rightarrow \text{biggest } \binom{n}{k} (= \binom{n}{n/2}) \geq \frac{1}{n+1} 2^n \quad \checkmark$$

(In fact, from **Stirling's approx.**  $n! \sim \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$ ,  
can show  $\binom{n}{n/2} \sim \sqrt{2/\pi n} 2^n$ .)

• histogram looks like it approaches a curve

In fact, letting  $n=2m$  for convenience, have

$$\binom{2m}{m-t} / \binom{2m}{m} \approx e^{-t^2/m}$$

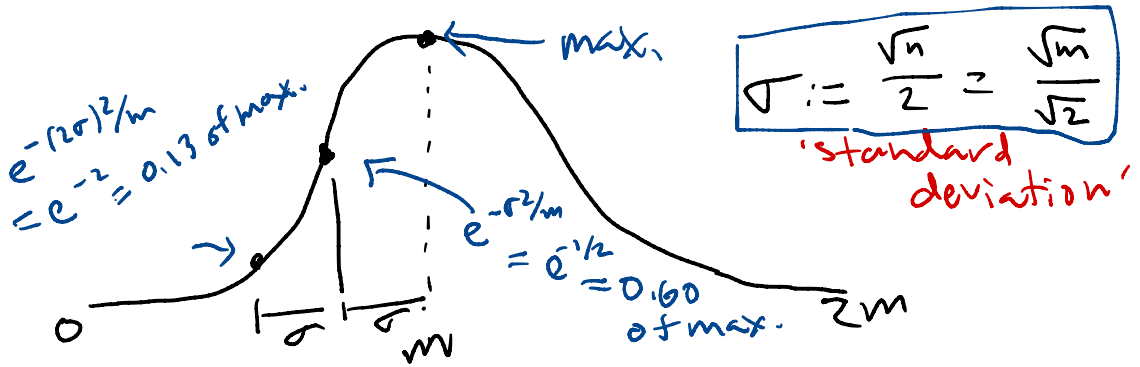
↑  
'Gaussian curve'

A.K.A. 'Normal curve'

A.K.A. 'Bell curve'

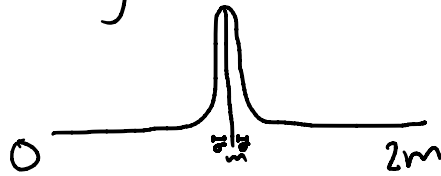
What does this mean?

Let's draw the curve to see...



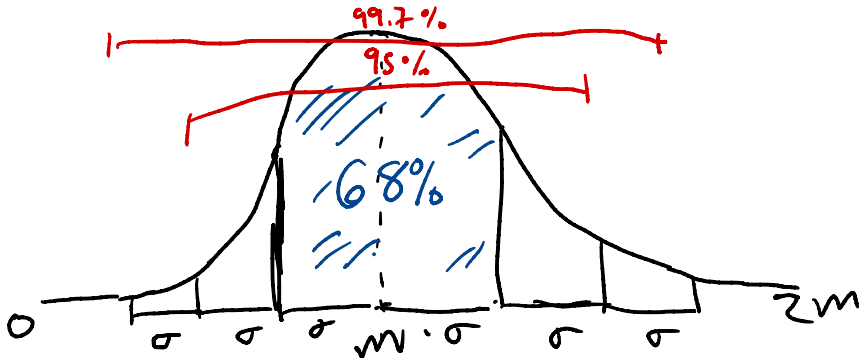
Note: picture is misleading since  $\sigma \ll 2m$

'Real picture' =



Upshot: Values  $\binom{n}{k}$  drop off rapidly from middle

Also, most of area under curve is in middle:



Precise lemmas from book are:

Lemma For  $0 \leq t \leq m$ ,

$$e^{-t^2/(m-t+1)} \leq \binom{2m}{m-t} / \binom{2m}{m} \leq e^{-t^2/(m+t)}$$

Lemma For  $0 \leq k \leq m$ , and  $c := \binom{2m}{k} / \binom{2m}{m}$ ,

$$\binom{2m}{0} + \binom{2m}{1} + \dots + \binom{2m}{k-1} < \frac{c}{2} \cdot 2^{2m}$$

total area under curve  $\rightarrow$  = sum of  $2m$ th row of  $\Delta$

e.g.,  $m = 500$

then  $\binom{1000}{448} / \binom{1000}{500} < 0.01$

Thus sum of 1st 447  $\binom{1000}{k} < 0.5\%$  of total sum

By symmetry, last 447  $\binom{1000}{k}$  also  $< 0.5\%$  total

So **middle 107 terms** account for  **$> 99\%$**  of sum of the 1000th row of Pascal's  $\Delta$ !

Pf of these lemmas: Skipped. Based on  $\binom{n}{k} = \frac{n!}{k!(n-k)!}$

+ manipulating inequalities, taking logarithms,

Stirling's approx., etc.  $\square$

Q: Why are we interested in these facts?

A: Basic Probability Theory!

$S$  = finite set = "sample space"

e.g.  $S = \{1, 2, 3, 4, 5, 6\}$  = outcomes of rolling a die

$S = \{H, T\}$  = flipping a coin

$S = \{HH, HT, TH, TT\}$  = flipping two coins

An event is any subset of  $S$ .

e.g. roll  $\geq 3 = \{3, 4, 5, 6\} \subseteq \{1, 2, 3, 4, 5, 6\}$

get one heads in two flips =  $\{HT, TH\} \subseteq \{HH, HT, TH, TT\}$

A probability distribution on  $S = \{s_1, s_2, \dots, s_n\}$

is a way of assigning nonnegative real numbers

$p(s_1), p(s_2), \dots, p(s_n)$  s.t.  $p(s_1) + \dots + p(s_n) = 1$ .

Commonly: uniform distribution  $p(s_i) = \frac{1}{n} \forall i$ .

(but can also allow a weighted die, etc...)

The probability of event  $A \subseteq S$  is  $\sum_{s_i \in A} P(s_i)$ .

If we have uniform distr., this is  $\frac{\#A}{\#S}$ .

e.g.  $\Pr(\text{roll} \geq 3) = \frac{\#\{3, 4, 5, 6\}}{\#\{1, 2, 3, 4, 5, 6\}} = \frac{4}{6} \checkmark$

Independence: Two events  $A, B \subseteq S$  are independent if  $\Pr(A \cap B) = \Pr(A) \cdot \Pr(B)$ .

Roughly,  $A + B$  are unrelated...

Do we see why uniform distr. on  $\{HH, HT, TH, TT\}$  = "flipping two independent fair coins"?

Q: What is the probability of getting exactly  $k$  heads when flipping  $n$  coins?

A:  $\frac{\binom{n}{k}}{2^n}$  = fraction of area under  <sup>$n^{\text{th}}$  row</sup> Pascal's  $\Delta$  curve at position  $k$



Let's see this in action w/ Galton board...

Q: If we do 1000 coin flips, what fraction of heads should we expect?

Thm (Law of Large numbers)

For any  $\epsilon > 0$ ,

$P_r$  (fraction of heads in  $n$  coin flips)  $\rightarrow 1$   
is between  $\frac{1}{2} - \epsilon$  and  $\frac{1}{2} + \epsilon$

as  $n \rightarrow \infty$ .

i.e., in 1000 coin flips, should expect very close to 50% heads!

Pf: Recall picture of  $n^{\text{th}}$  row of Pascal's  $\Delta$ :



All the mass is very close to middle  
 $= 50\%$  heads  $\square$

A more precise result called the Central limit theorem says that

as  $n \rightarrow \infty$ , histogram of

$$\sqrt{n} \left( \frac{1}{n} (\# \text{ of heads in } n \text{ flips}) - \frac{1}{2} \right) \rightarrow \frac{2}{\sqrt{2\pi}} e^{-2x^2}$$

↑  
rescale by  $\sqrt{n}$  to see 'fluctuations' from average

This just repeats what we saw earlier w/ Pascal's  $\Delta$ .

So what? Significance of LLN + CLT is that they apply not just to coin flips, but any time we take average of independent

random variables (e.g., dice rolling, error of scientific measurement, etc.) They explain:

- why the scientific method works
- why polling works, etc.

and why (in 'fairy tale land' at least) the

Gaussian curve emerges as a universal limit (e.g., human height distributions, etc.).

Now let's take a 5 min. break,  
and when we come back we  
can work on a worksheet on more  
combinatorial patterns in Pascal's  $\Delta$   
'in our breakout groups

(this worksheet is not really  
related to LLN/CLT...)