

Introduction to probability

Stat 134

FAII 2005 Berkeley

Lectures prepared by: Elchanan Mossel Yelena Shvets

Follows Jim Pitman's book: Probability Section 2.1

Toss a coin 100, what's the chance of 60 🐷?

P(a sequence with 60 \circledast) = (# of sequences with 60 \circledast)*P(particular sequence with 60 \circledast)= (# of sequences with 60 \circledast)* $\frac{1}{2}^{100}$

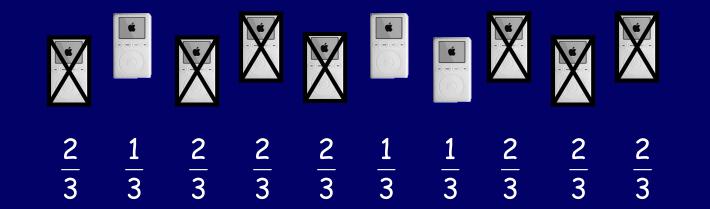
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hidden assumptions: n-independence; probabilities are fixed.

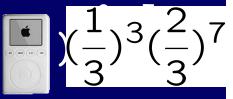




Each time we pull an item out of the hat it magically reappears. What's the chance of drawing 3 I-pods in 10 trials? $X = \{X > P\}$

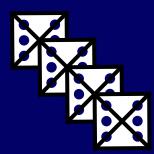


(3 📫 in 10 draws) = (# of sequences with 3 💻



Suppose we roll a die 4 times. What's the chance of k 2? Let $\mathbb{M} = \{ \mathbb{M} : \mathbb{N} : \mathbb{N} \}$.

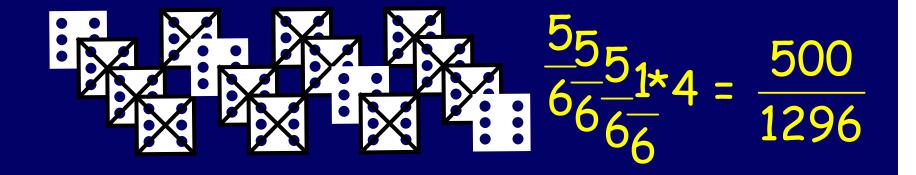
k=0



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Suppose we roll a die 4 times. What's the chance of k 2? Let $\mathbb{M} = \{\mathbb{M} : \mathbb{N} : \mathbb{N}\}$.

k=1



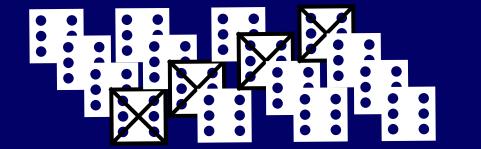
Suppose we roll a die 4 times. What's the chance of k 22 . Let $\mathbb{M} = \{\mathbb{M} : \mathbb{N} : \mathbb{N}\}$.

k=2

$\sum_{i=1}^{3} \frac{1}{666} + \frac{1}{666} + \frac{1}{666} + \frac{1}{666} + \frac{1}{666} + \frac{1}{1296} + \frac{1}{1296$

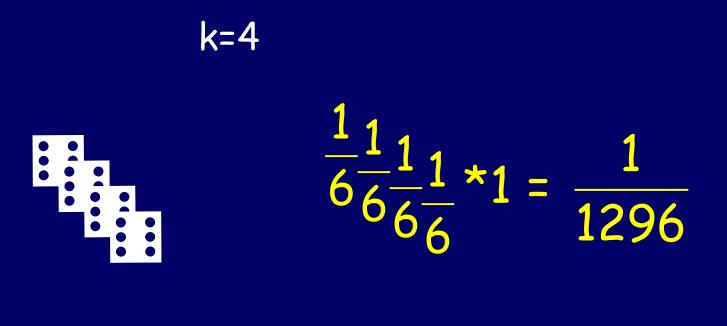
Suppose we roll a die 4 times. What's the chance of k 22 . Let $\mathbb{M} = \{\mathbb{M} : \mathbb{N} : \mathbb{N}\}$.

k=3

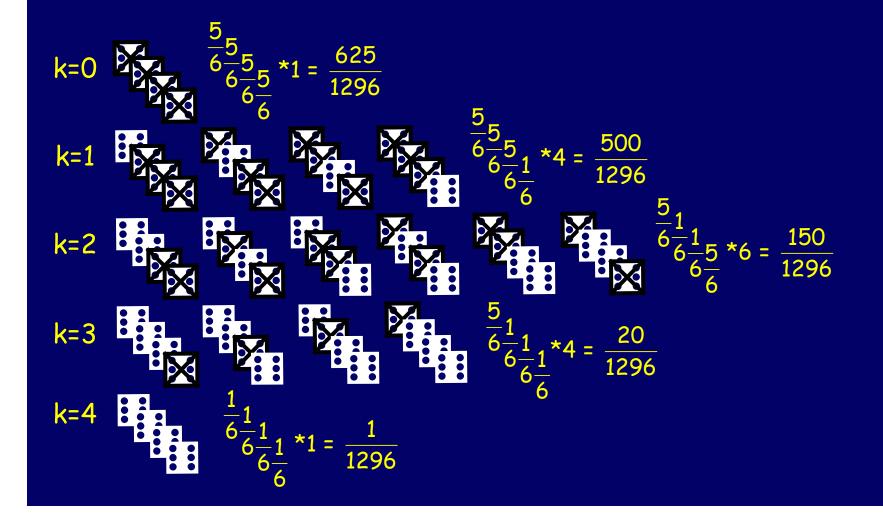


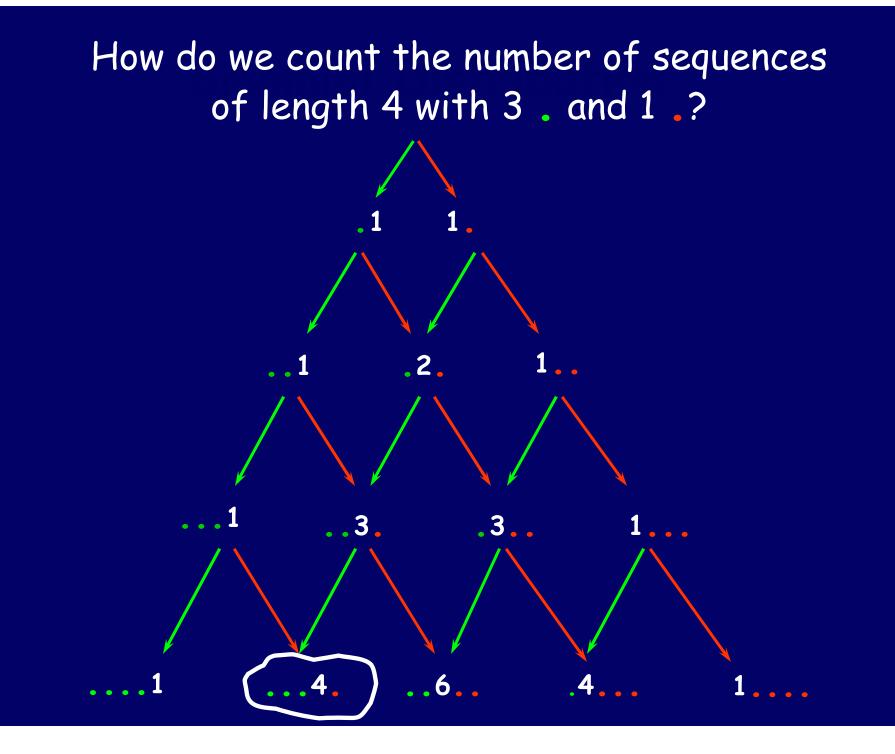
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Suppose we roll a die 4 times. What's the chance of k 2? Let $\mathbb{M} = \{ \mathbb{M} : \mathbb{N} : \mathbb{N} \}$.

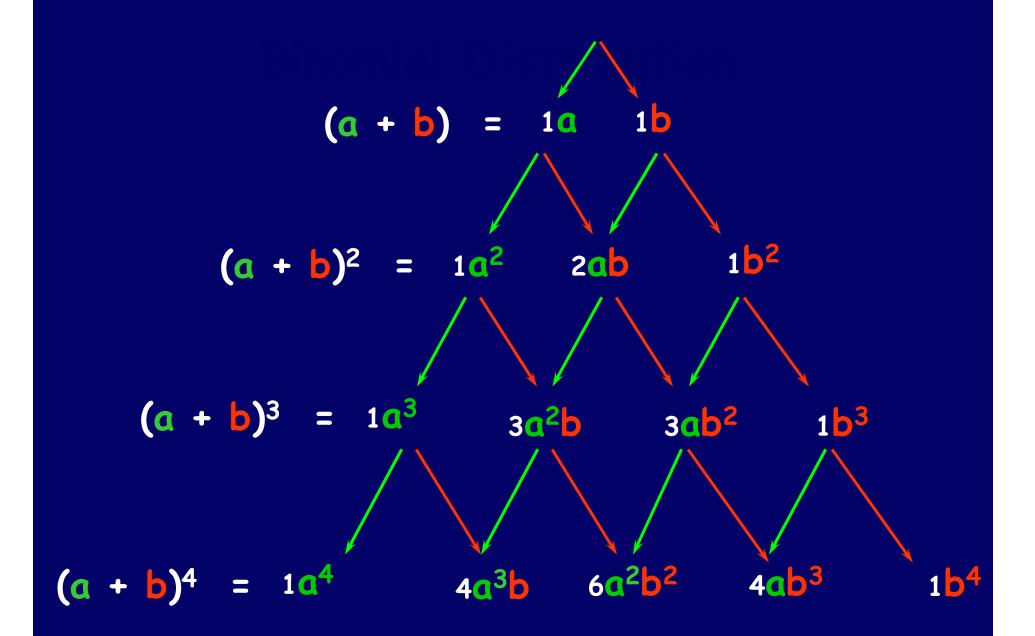


Suppose we roll a die 4 times. What's the chance of k 22 2 Let $\mathbb{M} = \{\mathbb{M} \cong \mathbb{M} : \mathbb{N} \}$.





This is the Pascal's triangle, which gives, as you may recall the binomial coefficients.



Newton's Binomial Theorem

$$(\mathbf{a} + \mathbf{b})^{n} = \sum_{k=0}^{n} \binom{\mathbf{n}}{\mathbf{k}} \mathbf{a}^{k} \mathbf{b}^{(n-k)}$$
$$\begin{pmatrix} \mathbf{n} \\ \mathbf{k} \end{pmatrix} = \frac{\mathbf{n}!}{\mathbf{k}!(\mathbf{n}-\mathbf{k})!}$$

Binomial Distribution

For *n* independent trials each with probability *p* of success and (1-p) of failure we have

$$P(\#successes=k) = \binom{n}{k} p^{k} (1-p)^{(n-k)}$$

This defines the *binomial(n,p)* distribution over the set of n+1 integers {0,1,...,n}.

Binomial Distribution

$$1 = \sum_{k=0}^{n} \binom{n}{k} p^{k} (1-p)^{(n-k)} = (p + (1-p))^{n}$$

This represents the chance that in n draws there was some number of successes between zero and n. A pair of coins will be tossed 5 times. Find the probability of getting \Im on k of the tosses, k = 0 to 5.



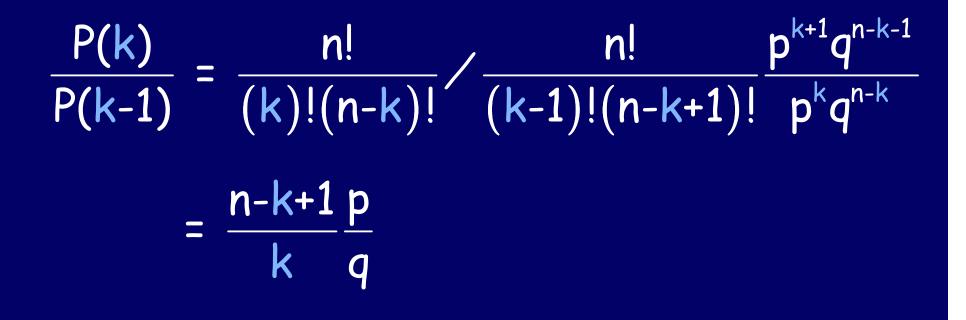
binomial(5,1/4)

A pair of coins will be tossed 5 times. Find the probability of getting \bigcirc on k of the tosses, k = 0 to 5.

$$\mathsf{P}\left(\# \overset{\circ}{\circ} = \mathsf{k}\right) = \binom{5}{\mathsf{k}} \left(\frac{1}{4}\right)^{\mathsf{k}} \left(\frac{3}{4}\right)^{5-\mathsf{k}}$$

To fill out the distribution table we could compute 6 quantities for k = 0,1,...5 separately, or use a trick.

Consecutive odds ratio relates P(k) and P(k-1).



Use consecutive odds ratio $\frac{P(k)}{P(k-1)} = \frac{n-k+1}{k} \frac{p}{q}$ to quickly fill out the distribution table for binomial(5,1/4):

$\frac{P(k)}{P(k-1)} = \frac{51}{13} + \frac{41}{23} + \frac{31}{33} + \frac{21}{43} + \frac{11}{53} $	k	0	1	2	3	4	5
$P(k) \begin{array}{c} 2\frac{3}{4}^{5} \\ 4\end{array} P(3)4\frac{5}{1}\frac{1}{3} P(2)6\frac{4}{1}\frac{1}{3} P(2)\frac{3}{3}\frac{1}{3} P(3)\frac{2}{4}\frac{1}{3} P(3)\frac{2}{4}\frac{1}{3} P(3)\frac{2}{3}\frac{1}{3} P(3)\frac{2}{4}\frac{1}{3} P(3)\frac{2}{3}\frac{1}{3} P(3)\frac{2}{3}\frac{1}$							
	P(k)	.2 <u>3</u> 5 47	P(3)40 13	$P(2) = \frac{41}{23}$	P.(28) 3 1 3 3	$P(3) \frac{2}{4} \frac{1}{3}$	RU(00)977 53

We can use this table to find the following conditional probability:

P(at least 3
$$\bigcirc$$
 | at least 1 \bigcirc in first 2 tosses)
= P(3 or more \bigcirc & 1 or 2 \bigcirc in first 2 tosses)
P(1 or 2 \bigcirc in first 2)

bin(5,1/4):

k	0	1	2	3	4	5
$\frac{P(k)}{P(k-1)}$		$\frac{51}{13}$	$\frac{41}{23}$	$\frac{3}{3}\frac{1}{3}$	$\frac{2}{4}\frac{1}{3}$	$\frac{1}{5}\frac{1}{3}$
P(k)	.237	.340	.264	.0879	.0146	.000977

bin(2,1/4):

k	0	1	2
$\frac{P(k)}{P(k-1)}$		$\frac{2}{1}\frac{1}{3}$	$\frac{1}{2}\frac{1}{3}$
P(k)	.563 43	P(3)751 13	P.(0) 2 3

How useful is the binomial formula? Try using your calculators to compute P(500 H in 1000 coin tosses) directly:

$P(500 \text{ in } 1000) = \frac{1000!}{500!500!} (\frac{1}{2})^{1000}$

Your calculator may return error when computing 1000!. This number is just too big to be stored.

The following is called the Stirling's approximation.

$$n! \approx \sqrt{2 \pi n} \left(\frac{n}{e}\right)^n$$

It is not very useful if applied directly : nⁿ is a very big number if n is 1000.

Stirling's formula:
$$n! \approx \sqrt{2 \pi n} \left(\frac{n}{e}\right)^n$$

 $P(500 \text{ in } 1000) = \frac{1000!}{500!500! 2} \frac{1}{2}^{1000}$
 $P(500 \text{ in } 1000) \approx \frac{1}{\sqrt{2 \pi}} \sqrt{\frac{1000}{500 \times 500}} \frac{\left(\frac{1000}{e}\right)^{1000}}{\left(\frac{500}{e}\right)^{500} \left(\frac{500}{e}\right)^{500}} \frac{1}{2}^{1000}}$
 $P(500 \text{ in } 1000) \approx \frac{1}{\sqrt{2 \pi}} \sqrt{\frac{2}{500}} \left(\frac{1000}{2 \times 500}\right)^{1000}}$
 $P(500 \text{ in } 1000) \approx \frac{1}{\sqrt{500 \pi}}$
 $P(500 \text{ in } 1000) \approx .0252313252$

Binomial Distribution

Toss coin 1000 times;

P(500 in 1000 250 in first 500)=

P(500 in 1000 & 250 in first 500)/P(250 in first 500)=

P(250 in first 500 & 250 in second 500)/P(250 in first 500)=

P(250 in first 500) P(250 in second 500)/ P(250 in first 500)=

 $P(250 \text{ in second } 500) = \frac{500!}{250!250!} \frac{1}{2}^{500} = \frac{1}{\sqrt{250\pi}} = 0.356824823$

Question: For a fair coin with $p = \frac{1}{2}$, what do we expect in 100 tosses?

Recall the frequency interpretation: $p \approx \#H/\#Trials$

So we expect about 50 H!

Expected value or Mean (µ) of a binomial(n,p) distribution

µ = #Trials × P(success)
 = np

Question: What is the most likely number of successes?

Mean seems a good guess.

Recall that P(50 in 100) $\approx \frac{1}{\sqrt{50\pi}} \approx .0797884561$

To see whether this is the most likely number of successes we need to compare this to P(k in 100) for every other k. The most likely number of successes is called the mode of a binomial distribution.

If we can show that for some m $P(1) \leq ... P(m-1) \leq P(m) \geq P(m+1) ... \geq P(n)$, then m would be the mode.

$P(k) \le P(l) \iff \frac{P(k)}{P(l)} \le 1,$

so we can use successive odds ratio:

$$\frac{P(k)}{P(k-1)} = \frac{n-k+1}{k} \frac{p}{q}$$

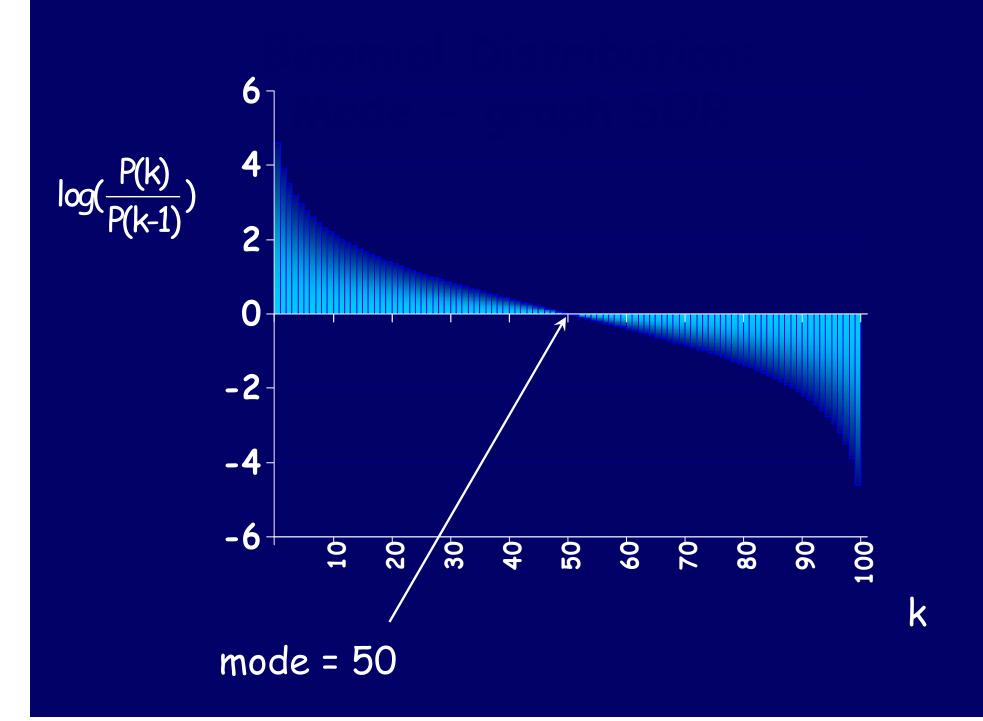
to determine the mode.

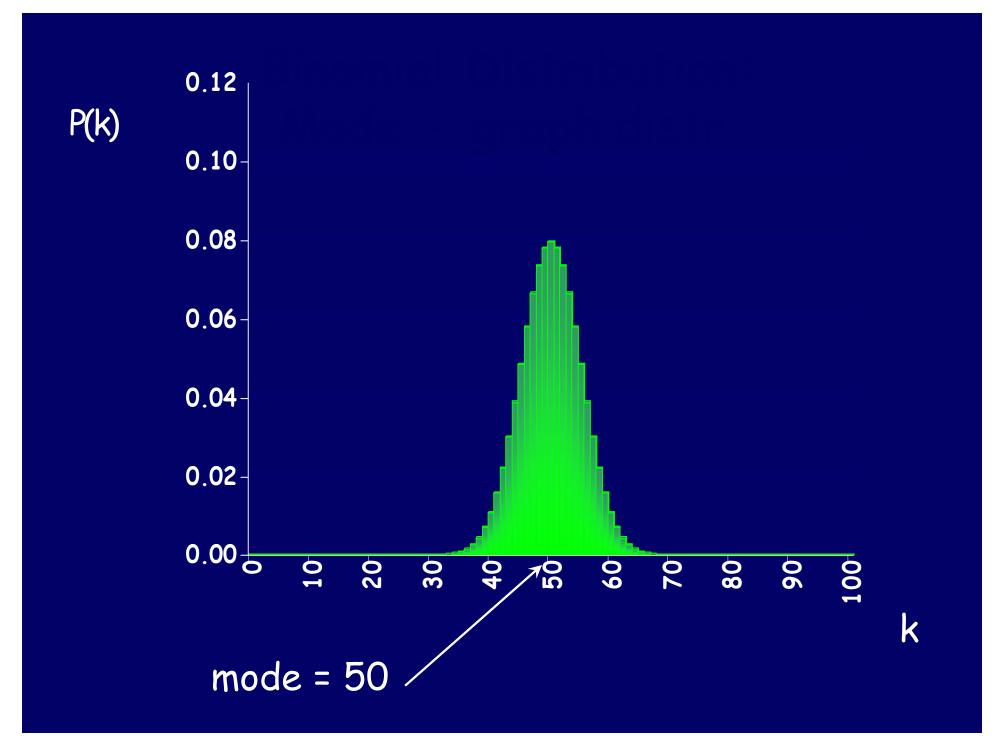
successive odds ratio: $\frac{P(k)}{P(k-1)} = \frac{n-k+1}{k} \frac{p}{1-p}$

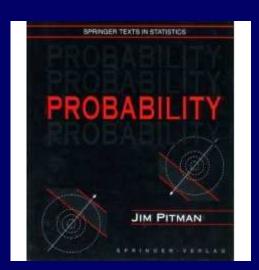
 $\begin{array}{l} P(k-1) > P(k) \iff \frac{k}{n-k+1} \frac{1-p}{p} > 1, \\ P(k-1) > P(k) \iff k(1-p) > (n-k+1)p \end{array}$

 $P(k-1) > P(k) \iff k > np+p;$

If we replace \leq with > the implications will still hold. So for m=[np+p] we get that P(m-1) \leq P(m) > P(m+1).







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