

The Binomial Distribution

n independent Bernoulli experiments

p, q

Each "success" with same probability p

"failure" with probability 1 - p

B_{p,n} - distribution of # successes

n independent coin flips

P(heads) = p

B_{p,n} - distribution of # heads

 $B_{n,p}$ more common Use $B_{p,n}$ Generalizes B_p , Natural for Poisson Binomial No confusion: $n \in \mathbb{N}$, $0 \le p \le 1$

Applications

Positive responses to a treatment

Faulty components

Rainy days in a month

Delayed flights

Smalln

n independent experiments

Success probability p Failure probability q = 1 - p

b_{p,n}(k) - probability of k successes

$$n = 0$$

$$n = 1$$

p+q=1

$$0 \le k \le n$$

$$n = 2$$

k	seq's	b _{p,2} (k)	
0	00	q ²	
1	01,10	2pq	
2	11	p ²	

$$p^2+2pq+q^2=(p+q)^2=1^2=1$$

General n and k

n

B_p experiments

successes 0 ≤ k ≤ n

 $b_{p,n}(k) = p(k successes)$

Every k-success sequence: n-k failures, probability pk-qn-k

 $\binom{n}{k}$ such sequences

$$= \binom{n}{k} p^k q^{n-k}$$

Distribution over n+1 values

$$0 \le k \le n$$

$$p(X = k) = b_{p,n}(k) = \binom{n}{k} p^k q^{n-k}$$

$$\sum_{k=0}^{n} b_{p,n}(k) = \sum_{k=0}^{n} \binom{n}{k} p^{k} q^{n-k}$$

$$= (p+q)^n$$

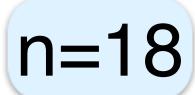
Binomial Theorem

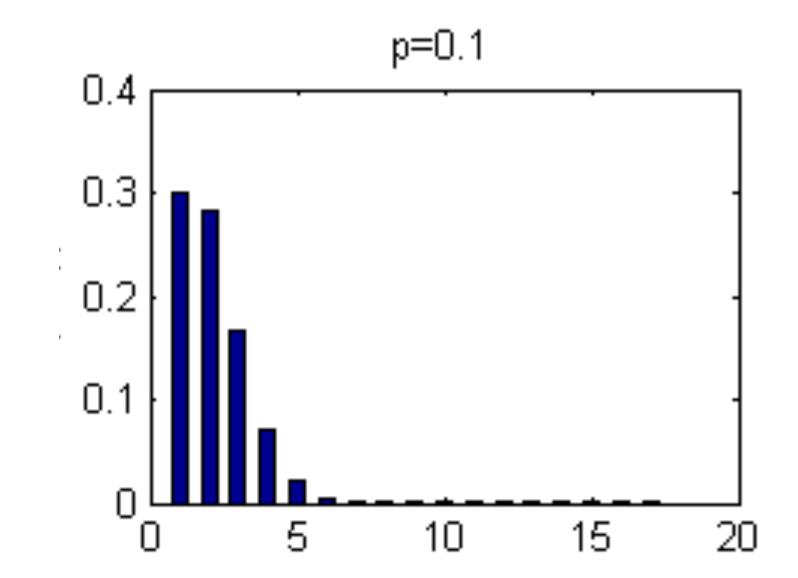
$$(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

$$= 1^n = 1$$

YES IT
ADDS!

Typical Distributions





Multiple Choice

Exam has 6 multiple-choice questions, each with 4 possible answers

Each question, student selects one of the 4 answers randomly

Passing: ≥ 4 correct answers P(passing) = ?

$$P(4) = {6 \choose 4} \cdot (\frac{1}{4})^4 \cdot (\frac{3}{4})^2 \approx 0.0329$$

$$P(5) = {6 \choose 5} \cdot (\frac{1}{4})^5 \cdot (\frac{3}{4})^1 \approx 0.00439$$

$$P(6) = {6 \choose 6} \cdot (\frac{1}{4})^6 \cdot (\frac{3}{4})^0 \approx 0.000244$$

$$P(\ge 4) = P(4) + P(5) + P(6) \approx 0.03759$$

Binomial as a Sum

B_{p,n} a sum of n B_p

$$X_1, \ldots, X_n \sim B_p \perp$$

$$X \stackrel{\text{def}}{=} \sum_{i=1}^{n} X_i$$

$$P(X = k) = P(\text{exactly k of } X_1, \dots, X_n \text{ are } 1) = {n \choose k} p^k q^{n-k} = b_{p,n}(k)$$

$$X \sim B_{p,n}$$

Apply to mean and variance

Mean and Variance

$$X \sim B_{p,n}$$

$$X = \sum_{i=1}^{n} X_i$$
 $X_1, ..., X_n \sim B_p \perp$

$$E(X) = E(\sum_{i=1}^{n} X_i) = \sum EX_i = \sum p = np$$

$$V(X) = V(\sum_{i=1}^{n} X_i) = \sum V(X_i) = \sum pq = npq$$

$$\sigma = \sqrt{npq}$$

Multiple Choice

Exam has 6 multiple-choice questions, each with 4 possible answers

For each question, student selects one of the 4 answers randomly

$$EX = np = 6 \cdot \frac{1}{4} = 1.5$$

Standard deviation

$$\sigma = \sqrt{npq} = \sqrt{6 \cdot \frac{1}{4} \cdot \frac{3}{4}} = \frac{\sqrt{18}}{4}$$

Why Vote

For simplicity odd # voters: 2n + 1

Each equally likely D or R

P(voter makes a difference) = P(other 2n voters equally split)

$$b_{p,n}(k) = \binom{n}{k} p^k q^{n-k}$$

$$= {2n \choose n} \frac{1}{2^n} \cdot \frac{1}{2^n}$$

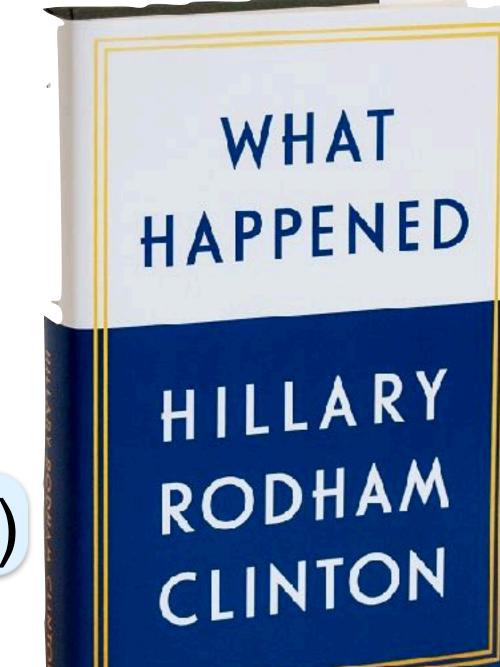
Stirling's Approximation

$$n! \sim \sqrt{2\pi n} (\frac{n}{e})^n$$

$$= \frac{(2n)!}{n! \cdot n! \cdot 2^n \cdot 2^n}$$

$$\approx \frac{\sqrt{2\pi \cdot 2n} (\frac{2n}{e})^{2n}}{(\sqrt{2\pi n} (\frac{n}{e})^n)^2 2^{2n}}$$

$$= \frac{1}{\sqrt{\pi n}}$$



Poisson Binomial

Generalizes the binomial distribution

n < 1	Binomial	B _{p,n}	For 1 ≤ i ≤ n	X _i ~ B _p	$X = \sum_{i=1}^{n} X_i$
1121	Poisson Binomial	PB _{p1} ,, p _n		Xi~Bpi	i=1

$$X_2 \sim B_{\frac{2}{3}}$$

X_1	X ₂	P	X
0	0	$3/4 \cdot 1/3 = 1/4$	0
0	1	$\frac{3}{4} \cdot \frac{2}{3} = \frac{1}{2}$	1
1	0	$\frac{1}{4} \cdot \frac{1}{3} = \frac{1}{12}$	1
1	1	$\frac{1}{4} \cdot \frac{2}{3} = \frac{1}{6}$	2

X	P(x)
0	1/4
1	7/12
2	1/6

Expectation and Variance

$$X \sim PB_{p_1, p_2, \dots, p_n} \qquad X = \sum_{i=1}^n X_i \qquad X_i \sim B_{p_i}$$

$$X = \sum_{i=1}^{n} X_i$$

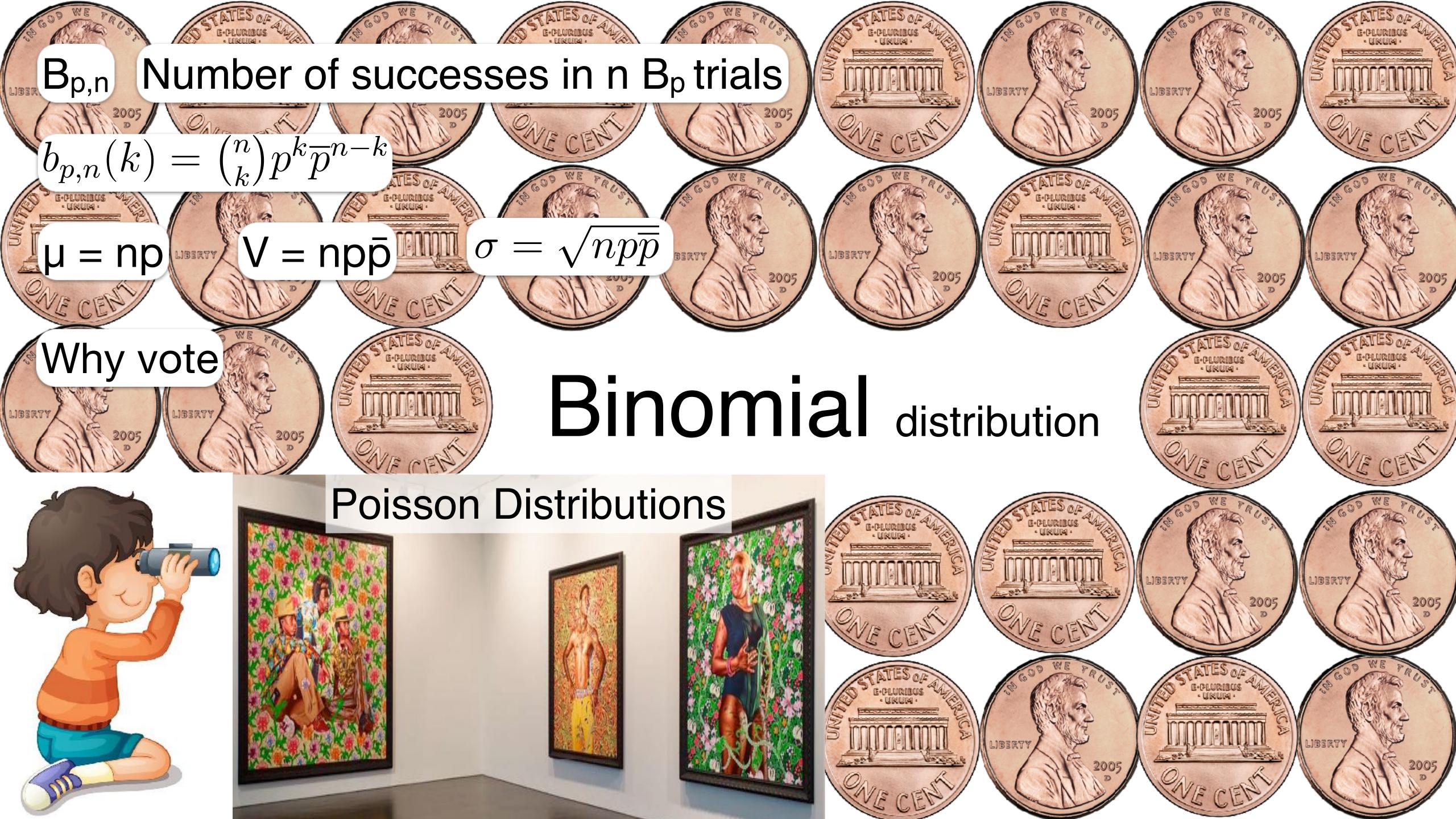
$$X_i \sim B_{p_i}$$

$$E(X) = E(\sum_{i=1}^{n} X_i) = \sum_{i=1}^{n} EX_i = \sum_{i=1}^{n} p_i$$

$$V(X) = V(\sum_{i=1}^{n} X_i) = \sum_{i=1}^{n} V(X_i) = \sum_{i=1}^{n} p_i (1 - p_i)$$

No closed form | Computationally

Homework



Coin Flips

Most basic convergence to average is B(p)

Flip n B(p) coins, average # 1's will approach np

Probability of a sequence with k 1's and n-k 0's is pkqn-k

Wolog assume p>0.5, then most likely is 1ⁿ

Yet by WLLN with probability \rightarrow 1 we see roughly pn 1's and qn 0's

Why do we observe these sequences and not the most likely ones?

Strength in #s. # sequences of a given composition increases near 1/2

pn balances # x probability.