

Debugging Intuition

- How to calculate the probability of at least k successes in n trials?

- X is number of successes in n trials each with probability p
- $P(X \geq k) =$

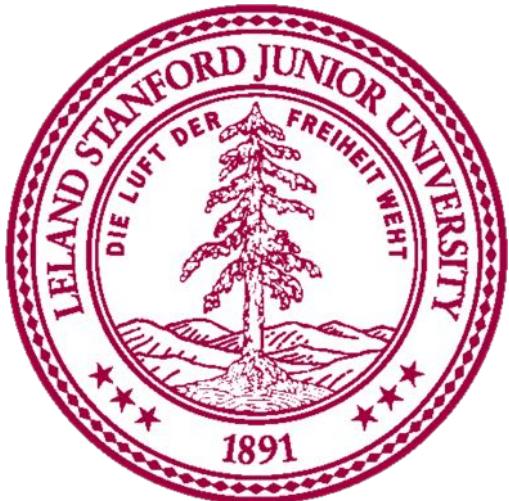
$$\binom{n}{k} p^k$$

ways to choose slots for success Don't care about the rest
Probability that each is success

First clue that something is wrong.
Think about $p = 1$

Not mutually exclusive...

Correct:
$$P(X \geq k) = \sum_{i=k}^n \binom{n}{i} p^i (1-p)^{n-i}$$



Variance

Chris Piech

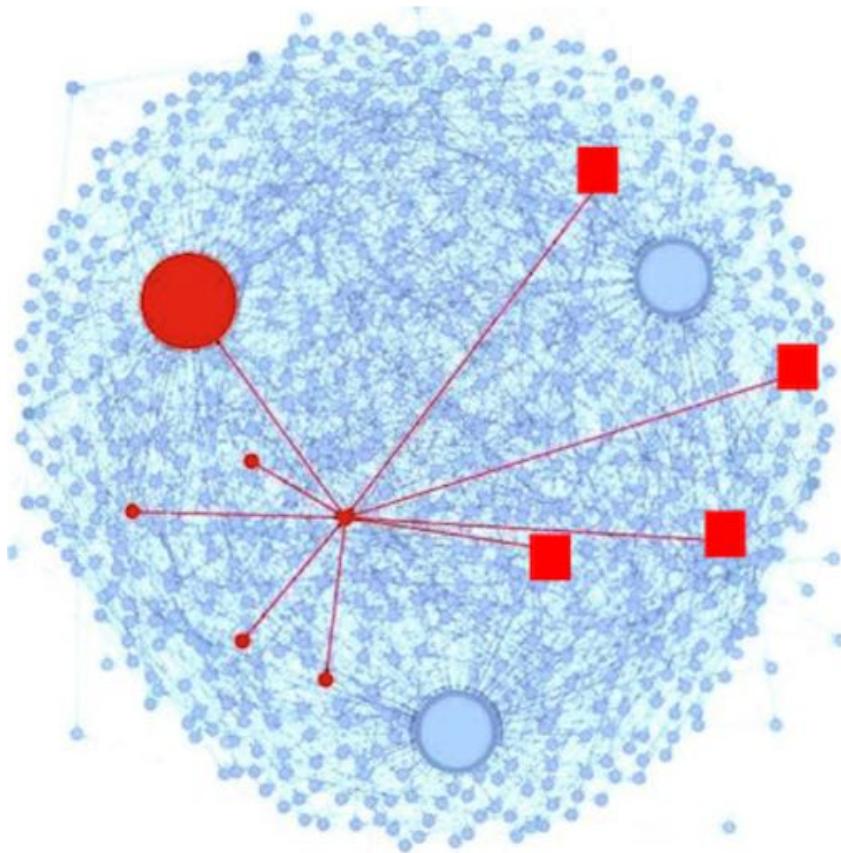
CS109, Stanford University

Learning Goals

1. Be able to calculate variance for a random variable
2. Be able to recognize and use a Bernoulli Random Var
3. Be able to recognize and use a Binomial Random Var



Is Peer Grading Accurate Enough?



Peer Grading on Coursera
HCI.

31,067 peer grades for
3,607 students.



Review: Random Variables



A **random variable** takes on values probabilistically.

For example:
 X is the sum of two dice rolled.

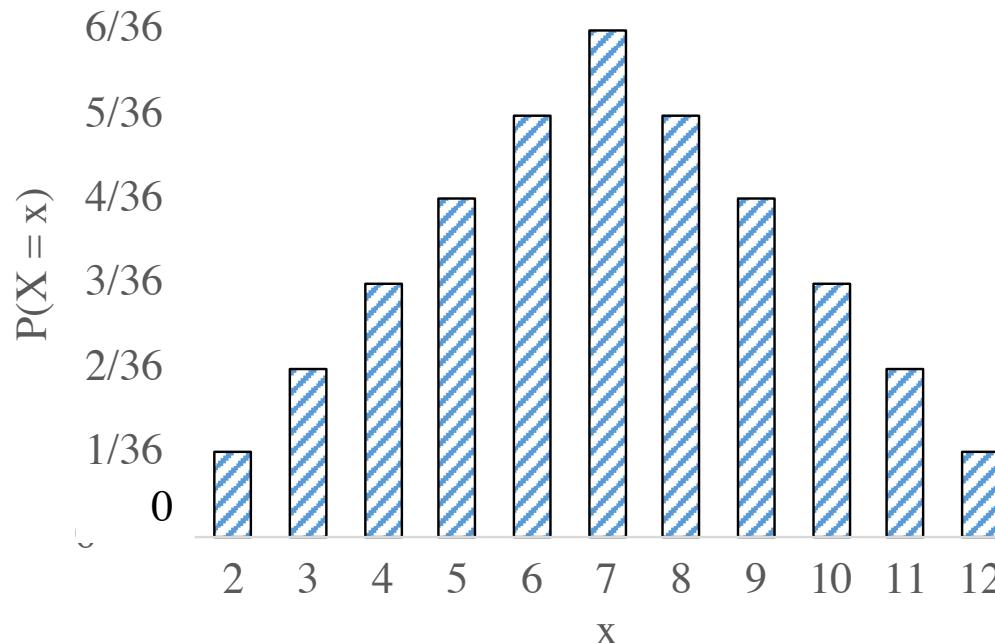
$$P(X = 2) = \frac{1}{36}$$

Review: Probability Mass Function



The **probability mass function** (PMF) of a random variable is a function from values of the variable to probabilities.

$$p_X(x) = P(X = x)$$

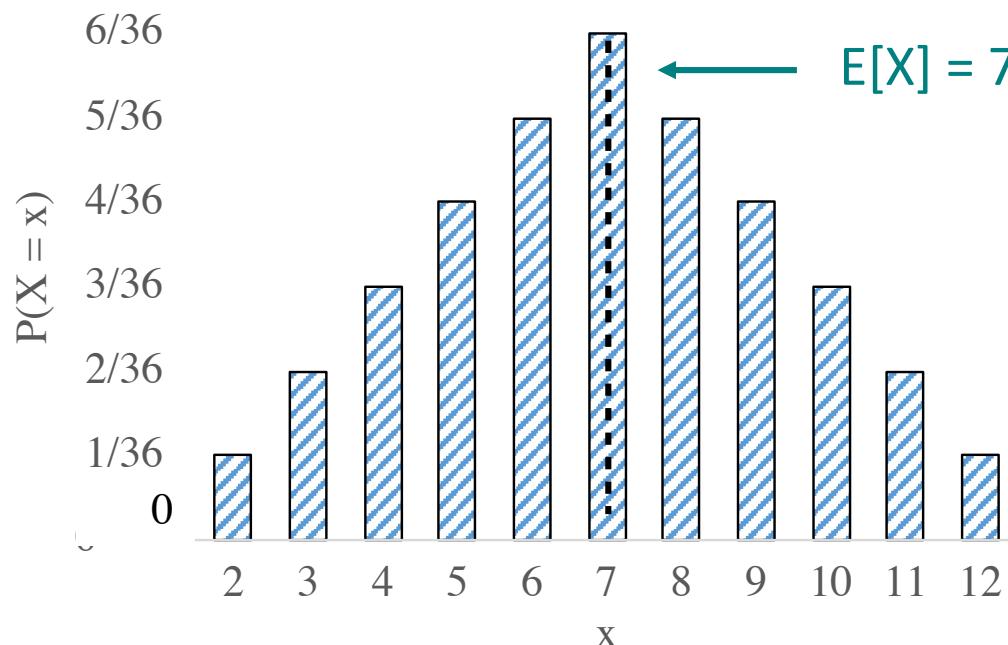


Review: Expectation



The **expectation** of a random variable is the “**average**” value of the variable (weighted by probability).

$$E[X] = \sum_{x:p(x)>0} p(x) \cdot x$$



Properties of Expectation

- **Linearity:**

$$E[aX + b] = aE[X] + b$$

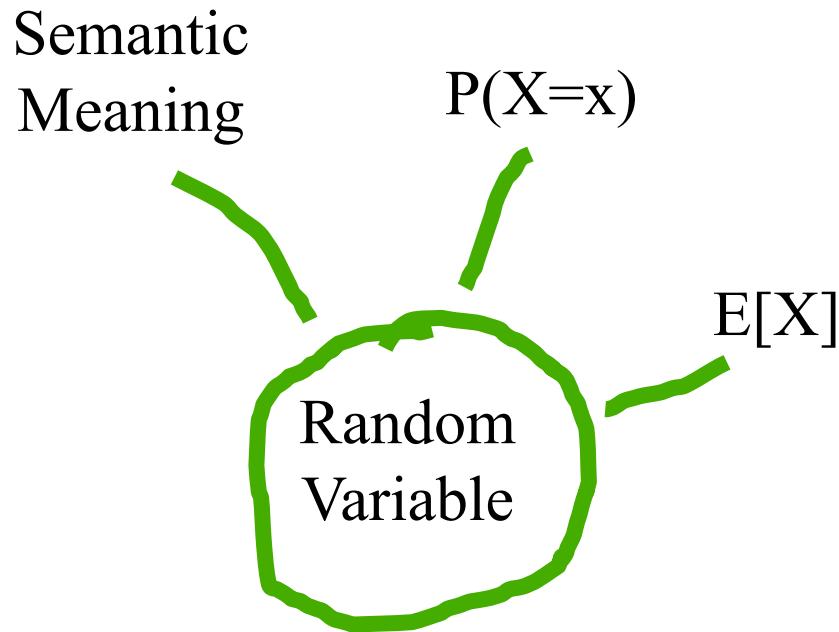
- **Expectation of a sum** is the sum of expectations

$$E[X + Y] = E[X] + E[Y]$$

- **Unconscious statistician:**

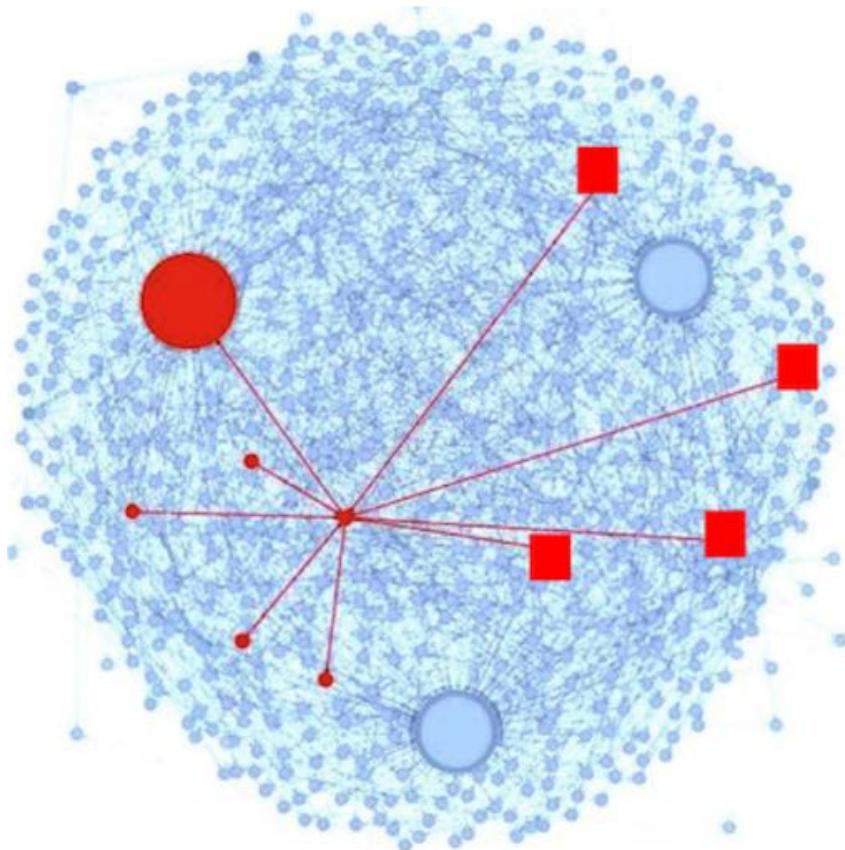
$$E[g(X)] = \sum_x g(x)p(x)$$

Fundamental Properties



Is $E[X]$ enough?

Intuition

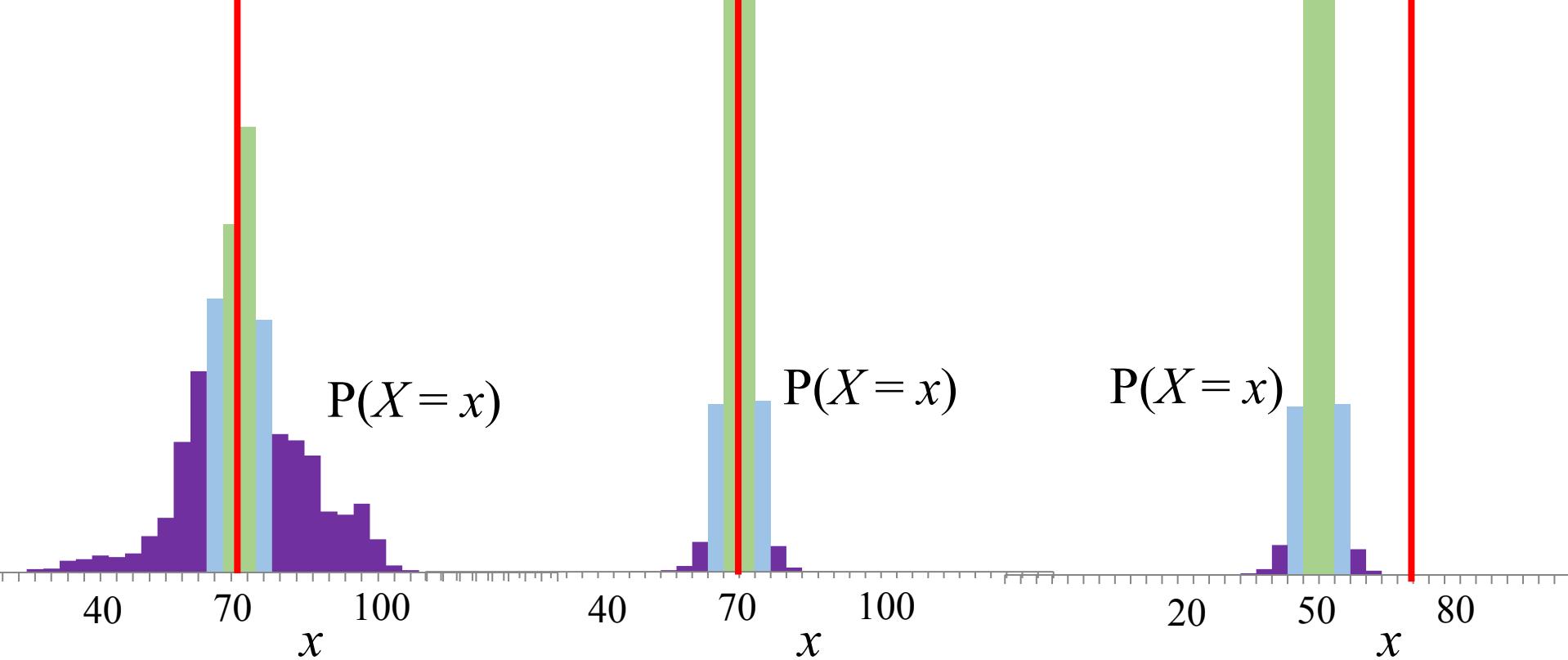


Peer Grading on Coursera
HCI.

31,067 peer grades for
3,607 students.

X is the score peer graders give to an assignment submission with true grade 70

True grade



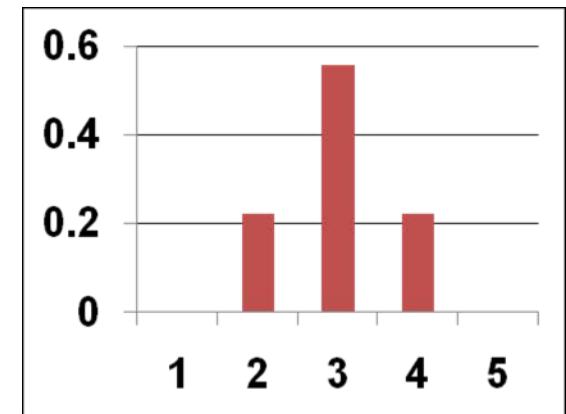
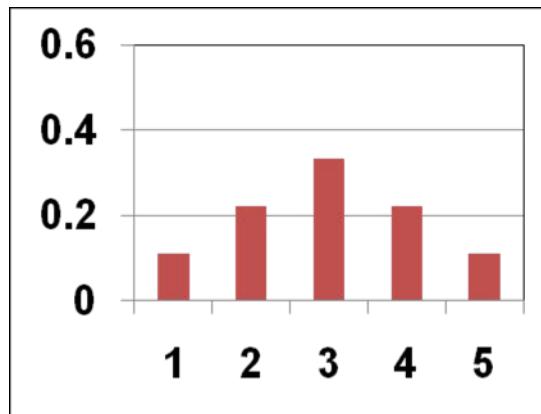
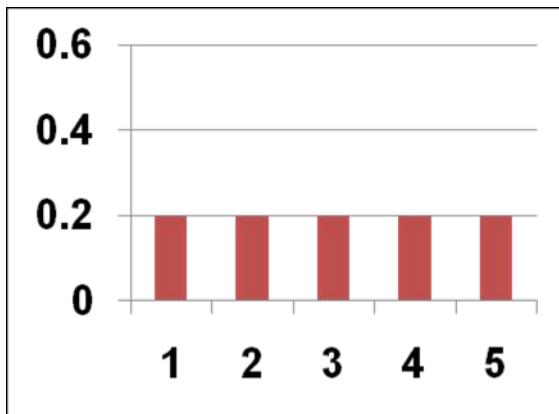
A

B

C

Variance

- Consider the following 3 distributions (PMFs)



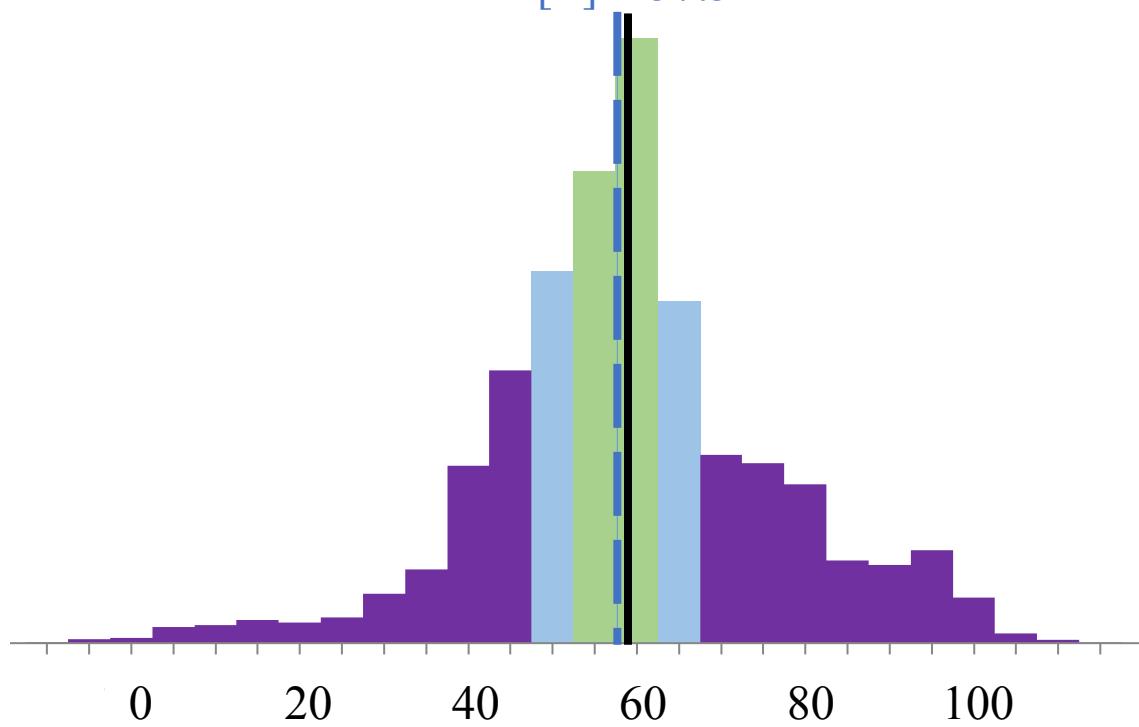
- All have the same expected value, $E[X] = 3$
- But “spread” in distributions is different
- Variance = a formal quantification of “spread”

Peer Grades in Coursera HCI

Let X be a random variable that represents a peer grade for an assignment that has a true grade of 58.

True grade = 58

$$E[X] = 57.5$$



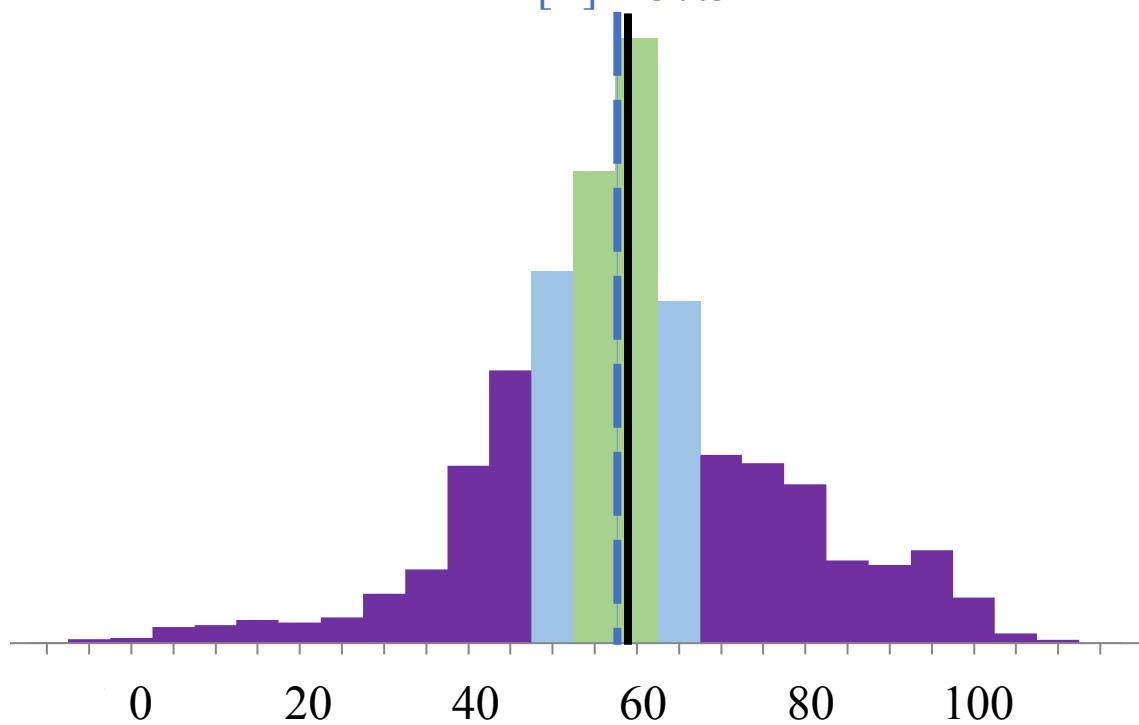
Peer Grades in Coursera HCI

Let X be a random variable that represents a peer grade

$$\text{Var}(X) = E[(X - \mu)^2]$$

True grade = 58

$$E[X] = 57.5$$



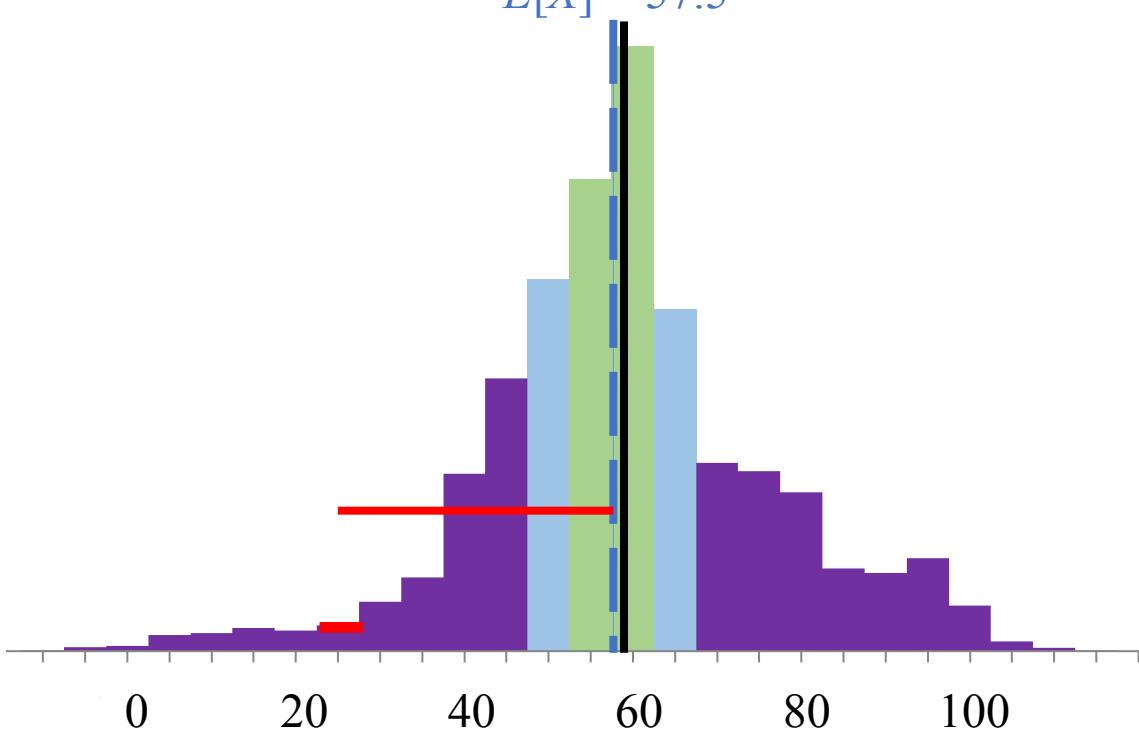
Peer Grades in Coursera HCI

Let X be a random variable that represents a peer grade

$$\text{Var}(X) = E[(X - \mu)^2]$$

True grade = 58
 $E[X] = 57.5$

X	$(X - \mu)^2$
25 points	1056 points ²



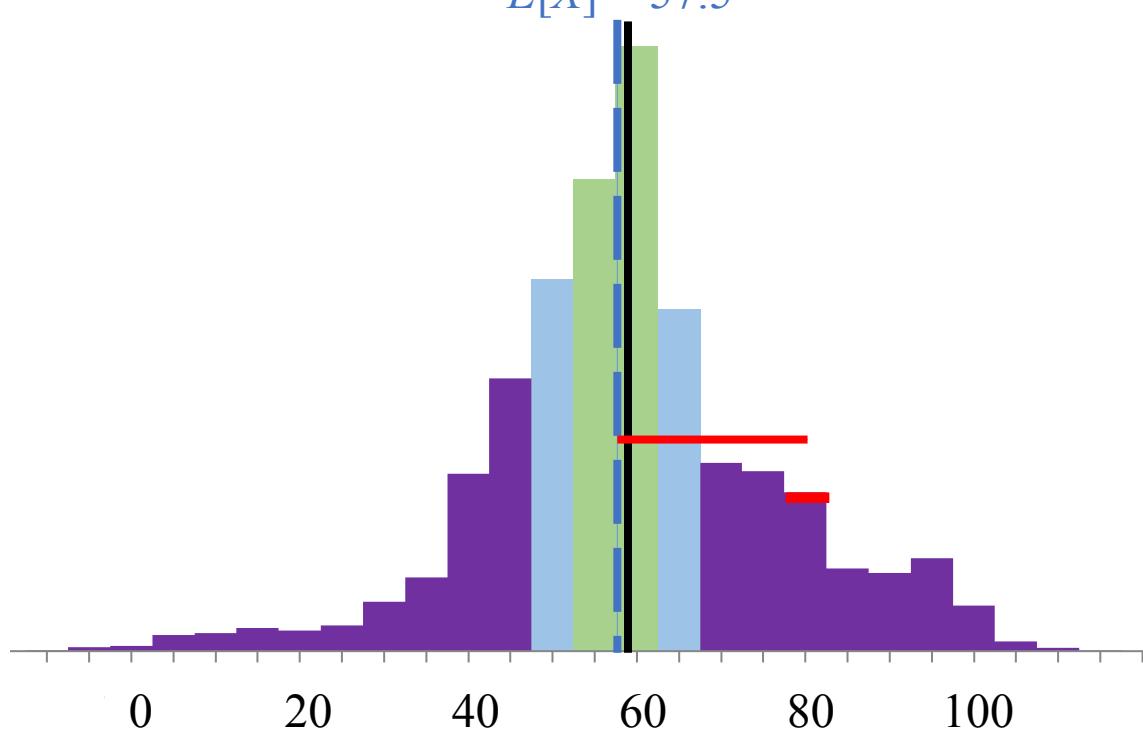
Peer Grades in Coursera HCI

Let X be a random variable that represents a peer grade

$$\text{Var}(X) = E[(X - \mu)^2]$$

True grade = 58

$$E[X] = 57.5$$



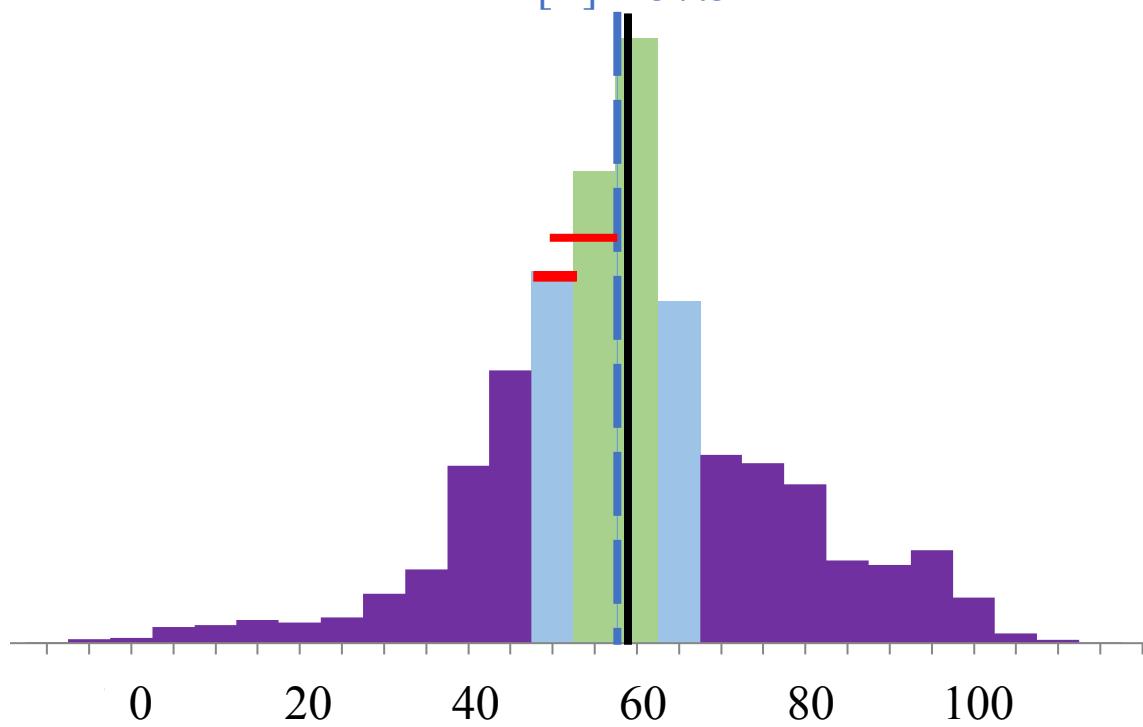
X	$(X - \mu)^2$
25 points	1056 points ²
80 points	506 points ²

Peer Grades in Coursera HCI

Let X be a random variable that represents a peer grade

$$\text{Var}(X) = E[(X - \mu)^2]$$

True grade = 58
 $E[X] = 57.5$



X	$(X - \mu)^2$
25 points	1056 points ²
80 points	506 points ²
50 points	56 points ²

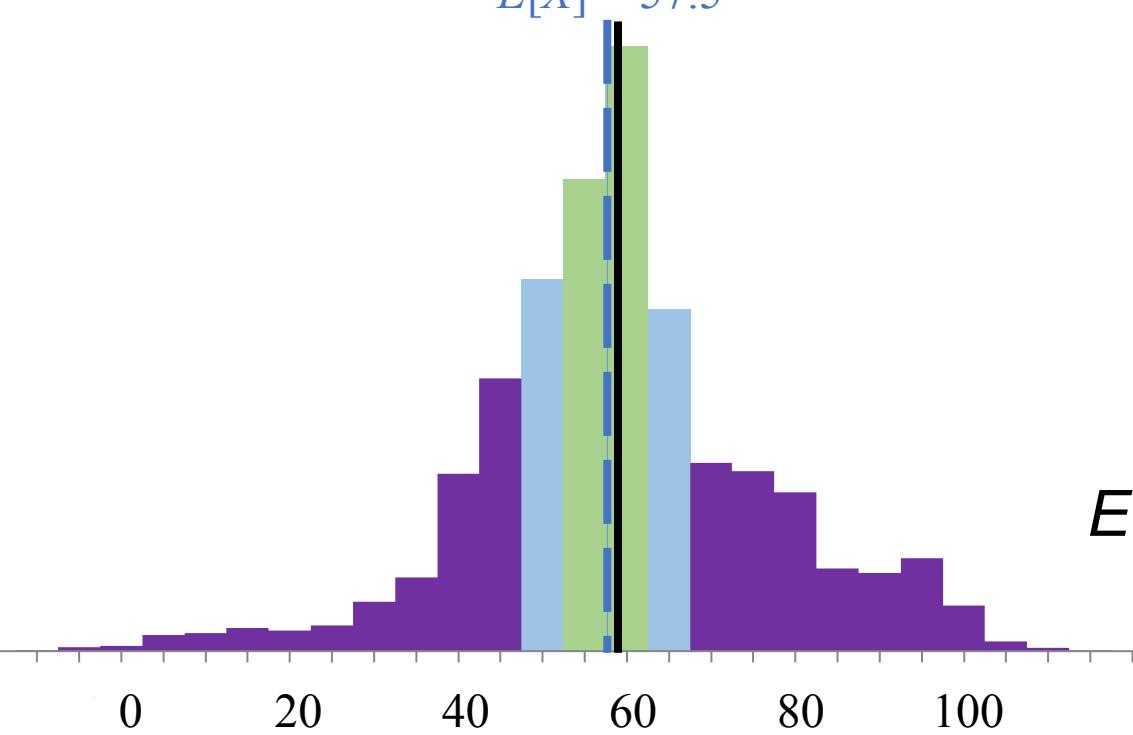
Peer Grades in Coursera HCI

Let X be a random variable that represents a peer grade

$$\text{Var}(X) = E[(X - \mu)^2]$$

True grade = 58

$$E[X] = 57.5$$



X	$(X - \mu)^2$
25 points	1056 points ²
80 points	506 points ²
50 points	56 points ²
...	

$$E [(X - \mu)^2] = 52 \text{ points}^2$$

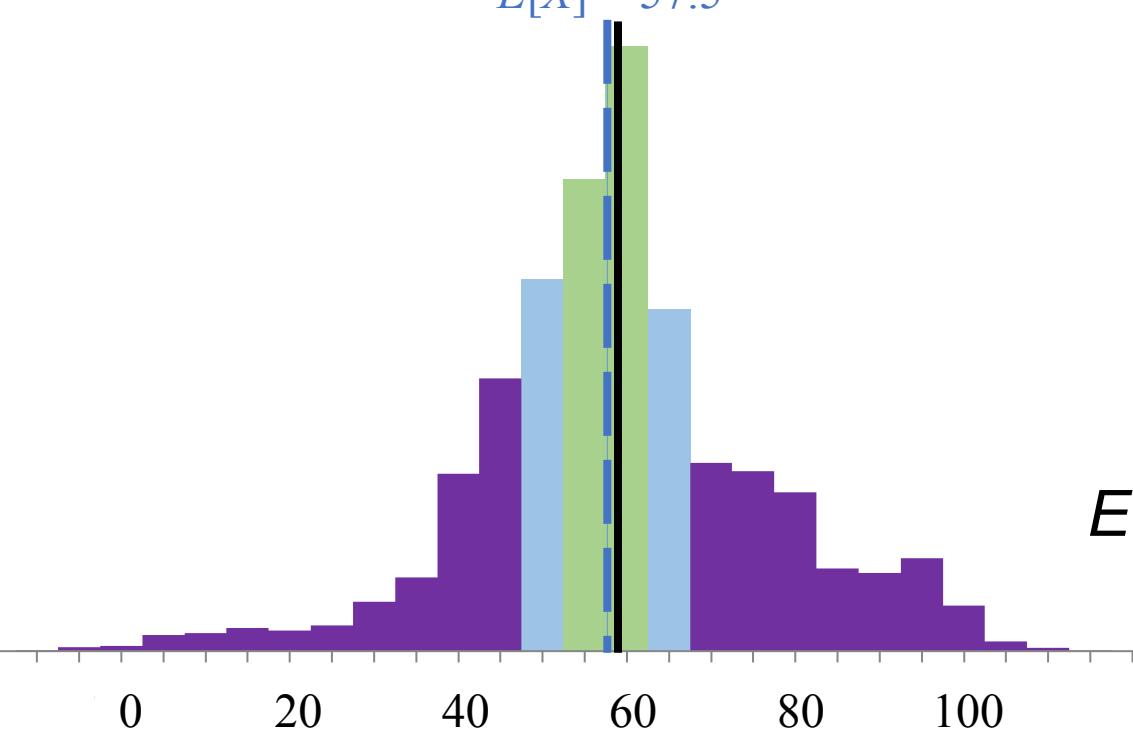
Peer Grades in Coursera HCI

Let X be a random variable that represents a peer grade

$$\text{Var}(X) = E[(X - \mu)^2]$$

True grade = 58

$$E[X] = 57.5$$



X	$(X - \mu)^2$
25 points	1056 points ²
80 points	506 points ²
50 points	56 points ²
...	

$$E [(X - \mu)^2] = 52 \text{ points}^2$$

$$\text{Std}(X) = 7.2 \text{ points}$$

Variance

- If X is a random variable with mean μ then the **variance** of X , denoted $\text{Var}(X)$, is:

$$\text{Var}(X) = E[(X - \mu)^2]$$

- Note: $\text{Var}(X) \geq 0$
- Also known as the 2nd **Central Moment**, or square of the Standard Deviation

Computing Variance

$$\text{Var}(X) = E[(X - \mu)^2]$$

Recall: Unconscious statistician:

$$E[g(X)] = \sum_x g(x)p(x)$$

$$\text{let } g(X) = (X - \mu)^2$$

Computing Variance

$$\begin{aligned}\text{Var}(X) &= E[(X - \mu)^2] \\ &= \sum_x (x - \mu)^2 p(x)\end{aligned}$$

Note: $\mu = E[X]$

$$\begin{aligned}&= \sum_x (x^2 - 2\mu x + \mu^2) p(x) \\ &= \sum_x x^2 p(x) - 2\mu \sum_x x p(x) + \mu^2 \sum_x p(x)\end{aligned}$$

$$= \boxed{E[X^2]} - 2\mu E[X] + \mu^2$$

$$= E[X^2] - 2\mu^2 + \mu^2$$

$$= E[X^2] - \mu^2$$

$$= \boxed{E[X^2] - (E[X])^2}$$

Ladies and gentlemen, please welcome the 2nd moment!

Variance of a 6 sided dice

- Let X = value on roll of 6 sided die
- Recall that $E[X] = 7/2$
- Compute $E[X^2]$

$$E[X^2] = (1^2)\frac{1}{6} + (2^2)\frac{1}{6} + (3^2)\frac{1}{6} + (4^2)\frac{1}{6} + (5^2)\frac{1}{6} + (6^2)\frac{1}{6} = \frac{91}{6}$$

$$\text{Var}(X) = E[X^2] - (E[X])^2$$

$$= \frac{91}{6} - \left(\frac{7}{2}\right)^2 = \frac{35}{12}$$

Properties of Variance

- $\text{Var}(aX + b) = a^2 \text{Var}(X)$

- Proof:

$$\begin{aligned}\text{Var}(aX + b) &= E[(aX + b)^2] - (E[aX + b])^2 \\ &= E[a^2X^2 + 2abX + b^2] - (aE[X] + b)^2 \\ &= a^2E[X^2] + 2abE[X] + b^2 - (a^2(E[X])^2 + 2abE[X] + b^2) \\ &= a^2E[X^2] - a^2(E[X])^2 = a^2(E[X^2] - (E[X])^2) \\ &= a^2\text{Var}(X)\end{aligned}$$

- Standard Deviation of X , denoted $\text{SD}(X)$, is:

$$\text{SD}(X) = \sqrt{\text{Var}(X)}$$

- $\text{Var}(X)$ is in units of X^2
 - $\text{SD}(X)$ is in same units as X

Fundamental Properties

Semantic
Meaning

$$P(X=x)$$

Random
Variable

$$E[X]$$

$$\text{Var}(X)$$

$$\text{Std}(X)$$

$$E[X^2]$$

Lots of fun with Random Variables

Classics



Jacob Bernoulli

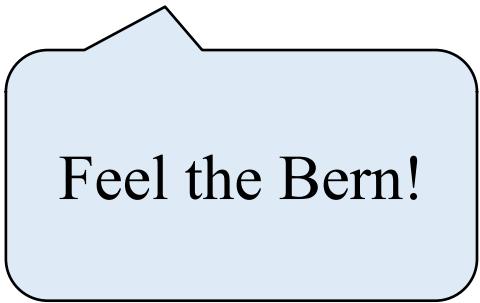
- Jacob Bernoulli (1654-1705), also known as “James”, was a Swiss mathematician



- One of many mathematicians in Bernoulli family
- The Bernoulli Random Variable is named for him
- He is my *academic* great¹²-grandfather
- Ice Cube at a renaissance fair?

Bernoulli Random Variable

- Experiment results in “Success” or “Failure”
 - X is random **indicator** variable ($1 = \text{success}$, $0 = \text{failure}$)
 - $P(X = 1) = p$ $P(X = 0) = 1 - p$
 - X is a Bernoulli Random Variable: $X \sim \text{Ber}(p)$
 - $E[X] = p$
 - $\text{Var}(X) = p(1 - p)$
- Examples
 - coin flip
 - random binary digit
 - whether a disk drive crashed
 - whether someone likes a netflix movie



Feel the Bern!

Does a Program Crash?



Run a program, crashes with probability $p = 0.1$,
works with probability $(1 - p)$

$\textcolor{blue}{X}$: 1 if program crashes

$$\text{P}(\textcolor{blue}{X} = 1) = p$$

$$\text{P}(\textcolor{blue}{X} = 0) = 1 - p$$

$$\textcolor{blue}{X} \sim \text{Ber}(p = 0.1)$$

Does a User Click an Ad?



Serve an ad, clicked with probability $p = 0.01$,
ignored with prob. $(1 - p)$

C: 1 if ad is clicked

$$P(C = 1) = p$$

$$P(C = 0) = 1 - p$$

$$C \sim \text{Ber}(p = 0.01)$$

More!

Binomial Random Variable

- Consider n **independent** trials of $\text{Ber}(p)$ rand. var.
 - Let X be the **number of successes** in n trials
 - X is a **Binomial** Random Variable: $X \sim \text{Bin}(n, p)$

$$P(X = i) = \binom{n}{i} p^i (1 - p)^{n-i} \text{ where } i \in \{0, 1, \dots, n\}$$

- Examples
 - # of heads in n coin flips
 - # of 1's in randomly generated length n bit string
 - # of disk drives crashed in 1000 computer cluster
 - Assuming disks crash independently

Bernoulli vs Binomial



Bernoulli is an indicator RV



Binomial is the sum of n Bernoullis

Three Coin Flips

- Three fair (“heads” with $p = 0.5$) coins are flipped
 - X is number of heads
 - $X \sim \text{Bin}(n = 3, p = 0.5)$

$$P(X = 0) = \binom{3}{0} p^0 (1-p)^3 = \frac{1}{8}$$

$$P(X = 1) = \binom{3}{1} p^1 (1-p)^2 = \frac{3}{8}$$

$$P(X = 2) = \binom{3}{2} p^2 (1-p)^1 = \frac{3}{8}$$

$$P(X = 3) = \binom{3}{3} p^3 (1-p)^0 = \frac{1}{8}$$

Properties of Bin(n, p)

Consider: $X \sim \text{Bin}(n, p)$

- $P(X = i) = \binom{n}{i} p^i (1 - p)^{n-i}$ where $i \in \{0, 1, \dots, n\}$
- $E[X] = np$
- $\text{Var}(X) = np(1 - p)$
- Note: $\text{Ber}(p) = \text{Bin}(1, p)$

Binomial distribution

Binomial distribution

From Wikipedia, the free encyclopedia

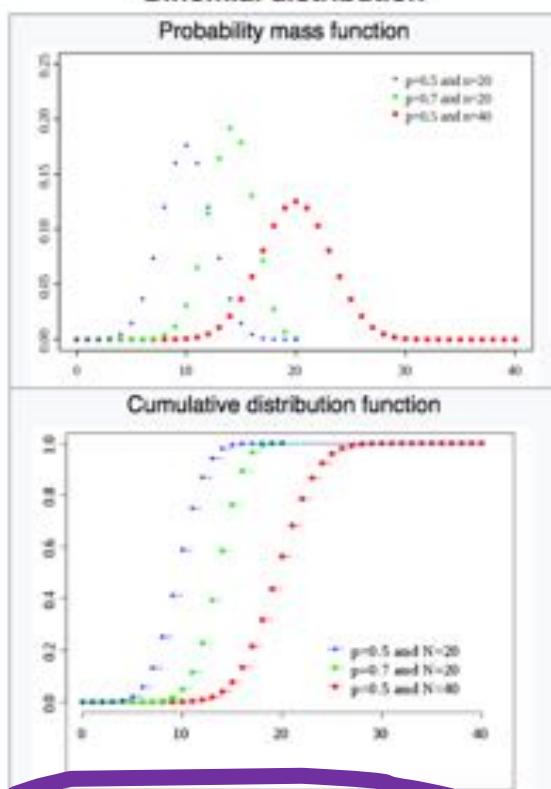
"Binomial model" redirects here. For the binomial model in finance, see [Binomial options pricing model](#).

Also, [Negative binomial distribution](#)

In probability theory and statistics, the **binomial distribution** with parameters n and p is the discrete probability distribution of n independent experiments, each asking a yes–no question, and each with its own boolean-valued outcome: a random variable that can have two possible outcomes (success/yes/true/one (with probability p) or failure/no/false/zero (with probability $q = 1 - p$). A single success/failure experiment is called a Bernoulli experiment and a sequence of outcomes is called a Bernoulli process; for a single trial, i.e., $n = 1$, the binomial distribution is the basis for the popular binomial test of statistical significance.

The binomial distribution is frequently used to model the number of successes in a sample of size n drawn with replacement from a population. If the sampling is carried out without replacement, the draws are not independent and so the resulting distribution is a hypergeometric distribution. However, if the sample size n is much larger than N , the binomial distribution remains a good approximation, and is widely used.

1 Specification
1.1 Probability mass function
1.2 Cumulative distribution function
2 Example
Mean
4 Variance
5 Mode
6 Median
7 Covariance between two binomials
8 Related distributions
8.1 Sums of binomials
8.2 Ratio of two binomial distributions
8.3 Conditional binomials
8.4 Bernoulli distribution
8.5 Poisson binomial distribution
8.6 Normal approximation
8.7 Poisson approximation
8.8 Limiting distributions
8.9 Beta distribution
9 Confidence intervals
9.1 Wald method
9.2 Agresti–Coull method ^[18]



Notation	$B(n, p)$
Parameters	$n \in \mathbb{N}_0$ — number of trials $p \in [0, 1]$ — success probability in each trial
Support	$k \in \{0, \dots, n\}$ — number of successes
pmf	$\binom{n}{k} p^k (1-p)^{n-k}$
CDF	$I_{1-p}(n - k, 1 + k)$
Mean	np
Median	$\lfloor np \rfloor$ or $\lceil np \rceil$
Mode	$\lfloor (n + 1)p \rfloor$ or $\lceil (n + 1)p \rceil - 1$
Variance	$np(1 - p)$
Skewness	$\frac{1 - 2p}{\sqrt{np(1 - p)}}$
Ex. kurtosis	$\frac{1 - 6p(1 - p)}{np(1 - p)}$
Entropy	$\frac{1}{p} \log_2 (2\pi e np(1 - p)) + O\left(\frac{1}{n}\right)$

I Really Want the Proof of Var :)

$$\begin{aligned} E(X^2) &= \sum_{k \geq 0}^n k^2 \binom{n}{k} p^k q^{n-k} \\ &= \sum_{k=0}^n k n \binom{n-1}{k-1} p^k q^{n-k} \\ &= np \sum_{k=1}^n k \binom{n-1}{k-1} p^{k-1} q^{(n-1)-(k-1)} \\ &= np \sum_{j=0}^m (j+1) \binom{m}{j} p^j q^{m-j} \\ &= np \left(\sum_{j=0}^m j \binom{m}{j} p^j q^{m-j} + \sum_{j=0}^m \binom{m}{j} p^j q^{m-j} \right) \\ &= np \left(\sum_{j=0}^m m \binom{m-1}{j-1} p^j q^{m-j} + \sum_{j=0}^m \binom{m}{j} p^j q^{m-j} \right) \\ &= np \left((n-1)p \sum_{j=1}^m \binom{m-1}{j-1} p^{j-1} q^{(m-1)-(j-1)} + \sum_{j=0}^m \binom{m}{j} p^j q^{m-j} \right) \\ &= np ((n-1)p(p+q)^{m-1} + (p+q)^m) \\ &= np ((n-1)p + 1) \\ &= n^2 p^2 + np(1-p) \end{aligned}$$

Definition of Binomial Distribution: $p + q = 1$

Factors of Binomial Coefficient: $k \binom{n}{k} = n \binom{n-1}{k-1}$

Change of limit: term is zero when $k-1=0$

putting $j=k-1, m=n-1$

splitting sum up into two

Factors of Binomial Coefficient: $j \binom{m}{j} = m \binom{m-1}{j-1}$

Change of limit: term is zero when $j-1=0$

Binomial Theorem

as $p+q=1$

by algebra

How Many Program Crashes?



n runs of program, each crashes with probability $p = 0.1$,
works with probability $(1 - p)$.

What is the probability of exactly 2 crashes with 100 users?

$\textcolor{blue}{H}$: number of crashes

$$\textcolor{blue}{H} \sim \text{Bin}(n = 100, p = 0.1)$$

$$\mathbf{P}(\textcolor{blue}{H} = k) = \binom{n}{k} (p)^k (1 - p)^{n-k}$$

$$P(H = 2) = \binom{100}{2} (0.1)^2 (0.9)^{98}$$

How Many Program Crashes?



n runs of program, each crashes with probability $p = 0.1$,
works with probability $(1 - p)$.

What is the probability of < 3 crashes with 100 users?

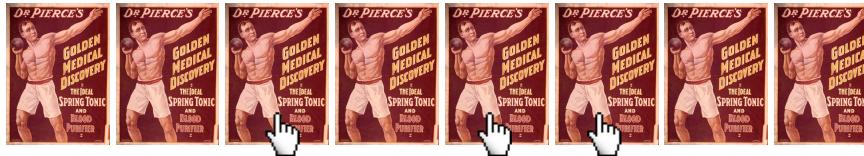
$\textcolor{blue}{H}$: number of crashes

$$\textcolor{blue}{H} \sim \text{Bin}(n = 100, p = 0.1)$$

$$\mathbf{P}(\textcolor{blue}{H} = k) = \binom{n}{k} (p)^k (1 - p)^{n-k}$$

$$P(H < 3) = \sum_{i=0}^2 \binom{100}{i} (0.1)^i (0.9)^{100-i}$$

How Many Ads Clicked?



1000 ads served, each clicked with $p = 0.01$, otherwise ignored.
Expectation and Standard deviation of number of ads clicked?

H : number of clicks

$$H \sim \text{Bin}(n = 1000, p = 0.01)$$

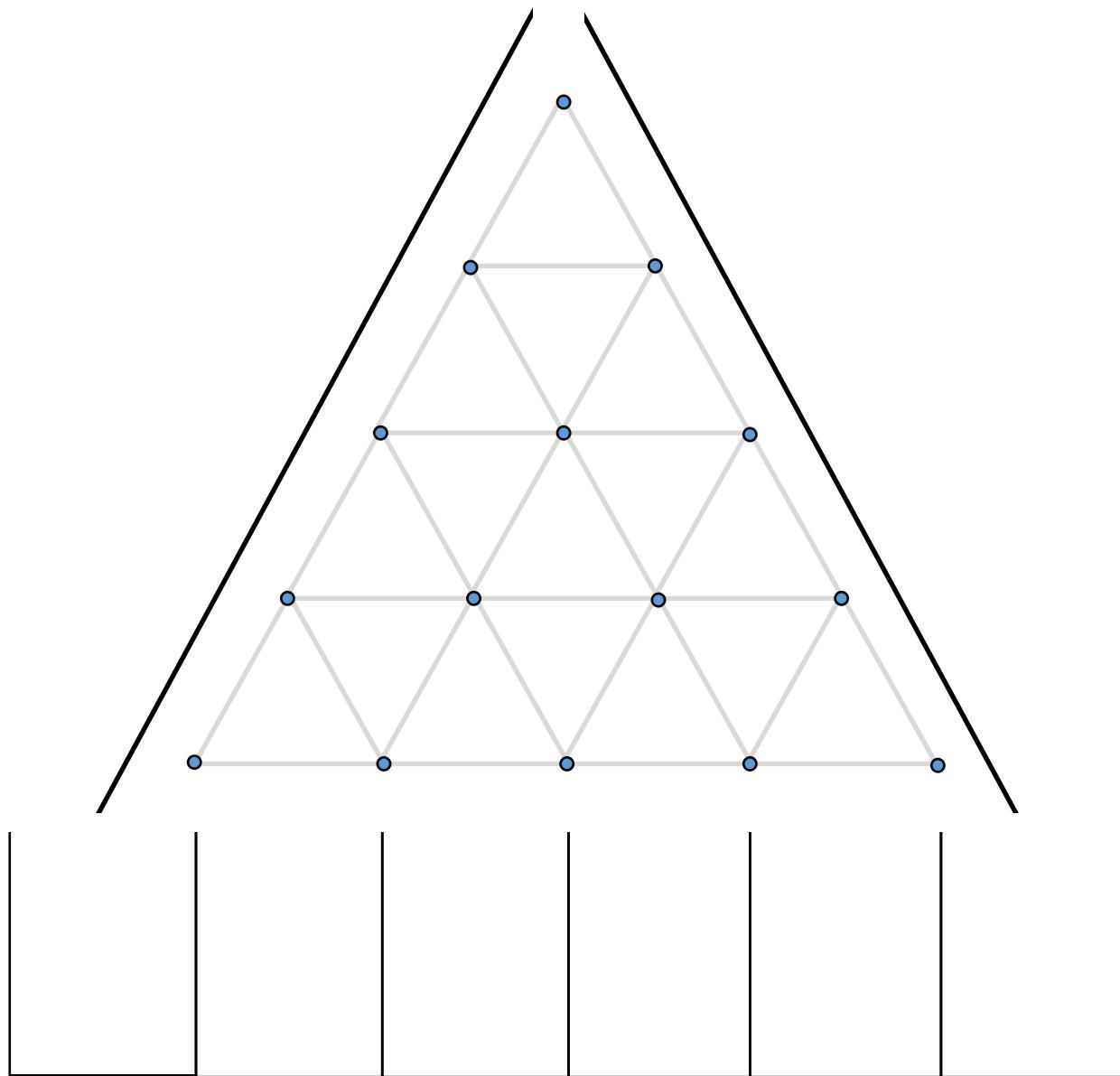
$$\mathbf{P}(H = k) = \binom{1000}{k} (0.01)^k (0.99)^{1000-k}$$

$$\mathbf{E}(H) = np = 10$$

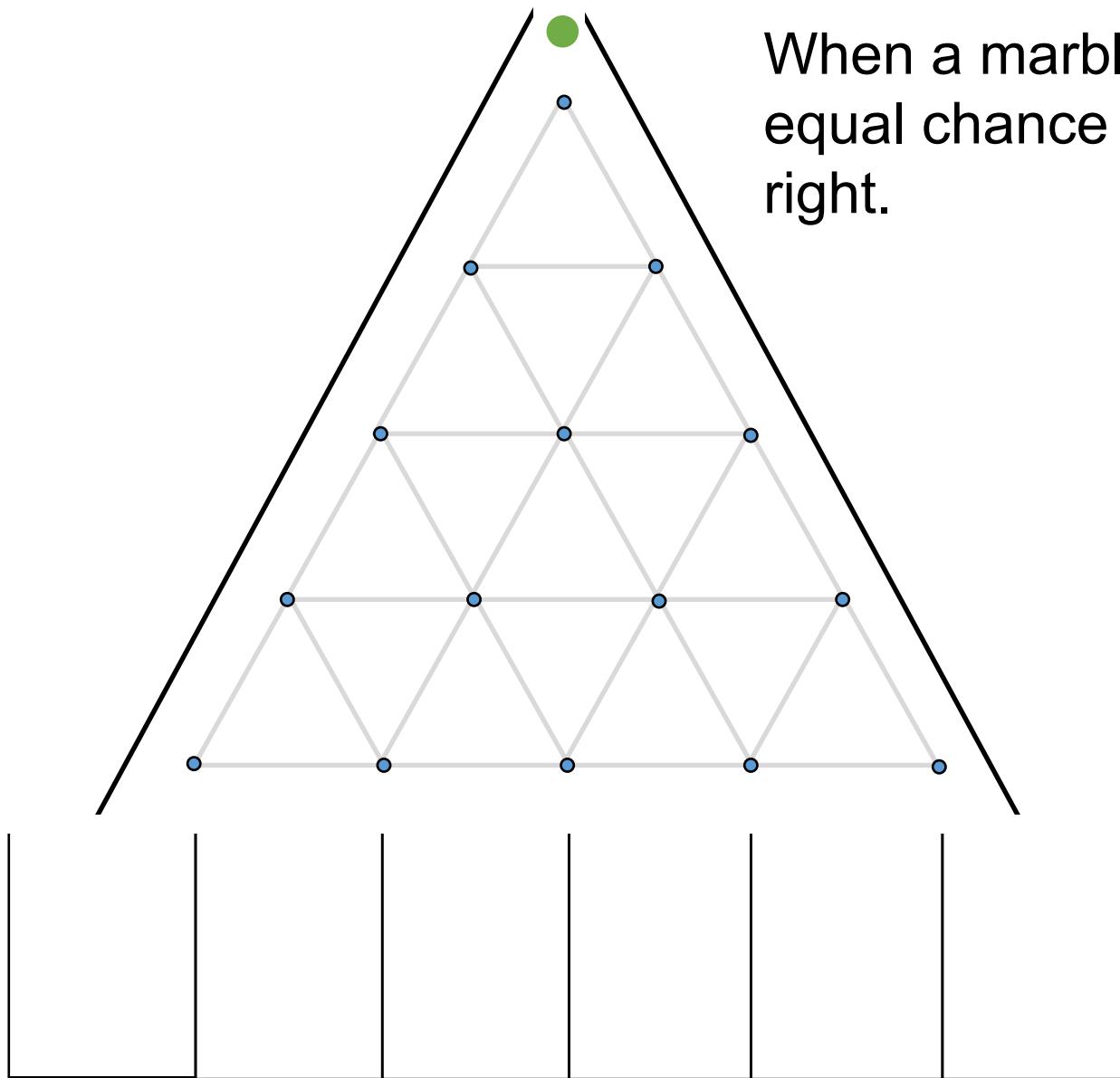
$$\mathbf{Var}(H) = np(1-p) = 9.9$$

$$\mathbf{Std}(H) = 3.15$$

Galton Board

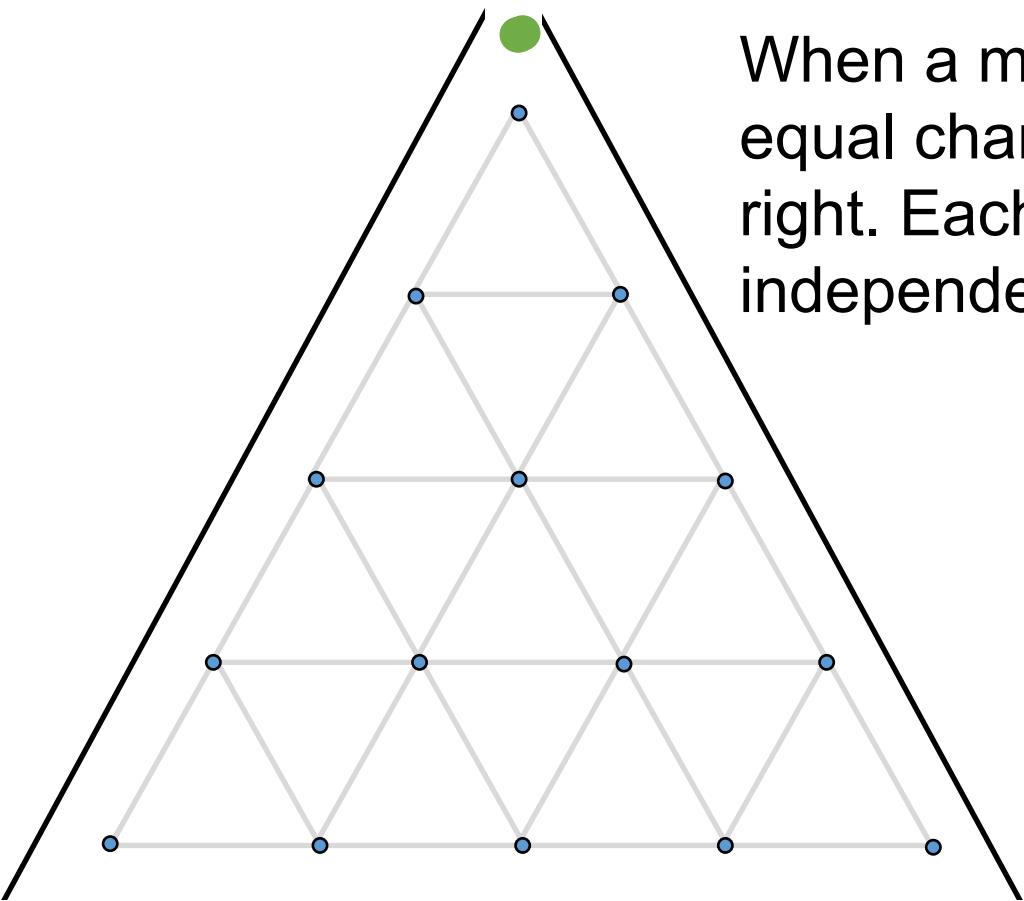


Galton Board



When a marble hits a pin, it has equal chance of going left or right.

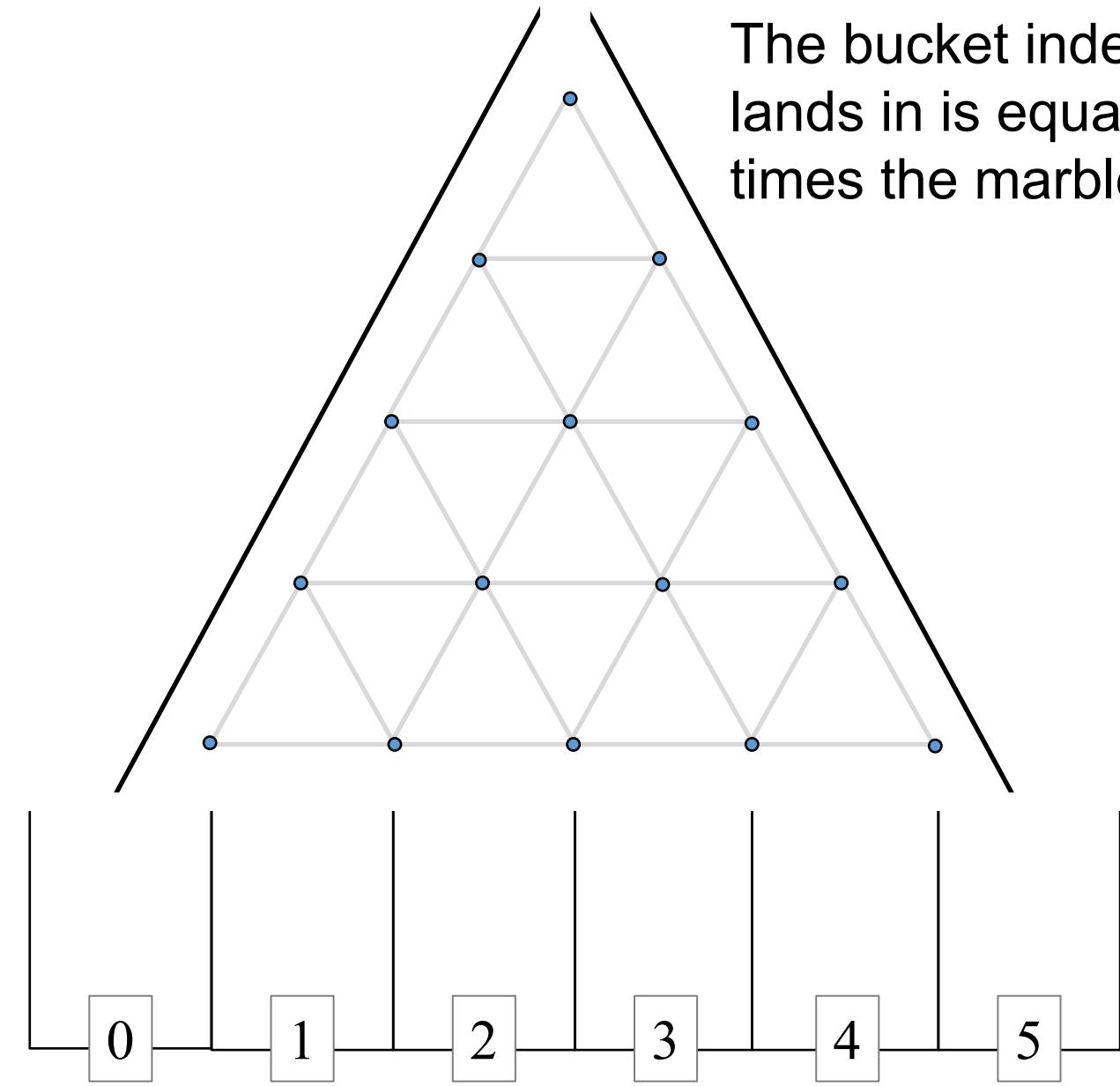
Galton Board



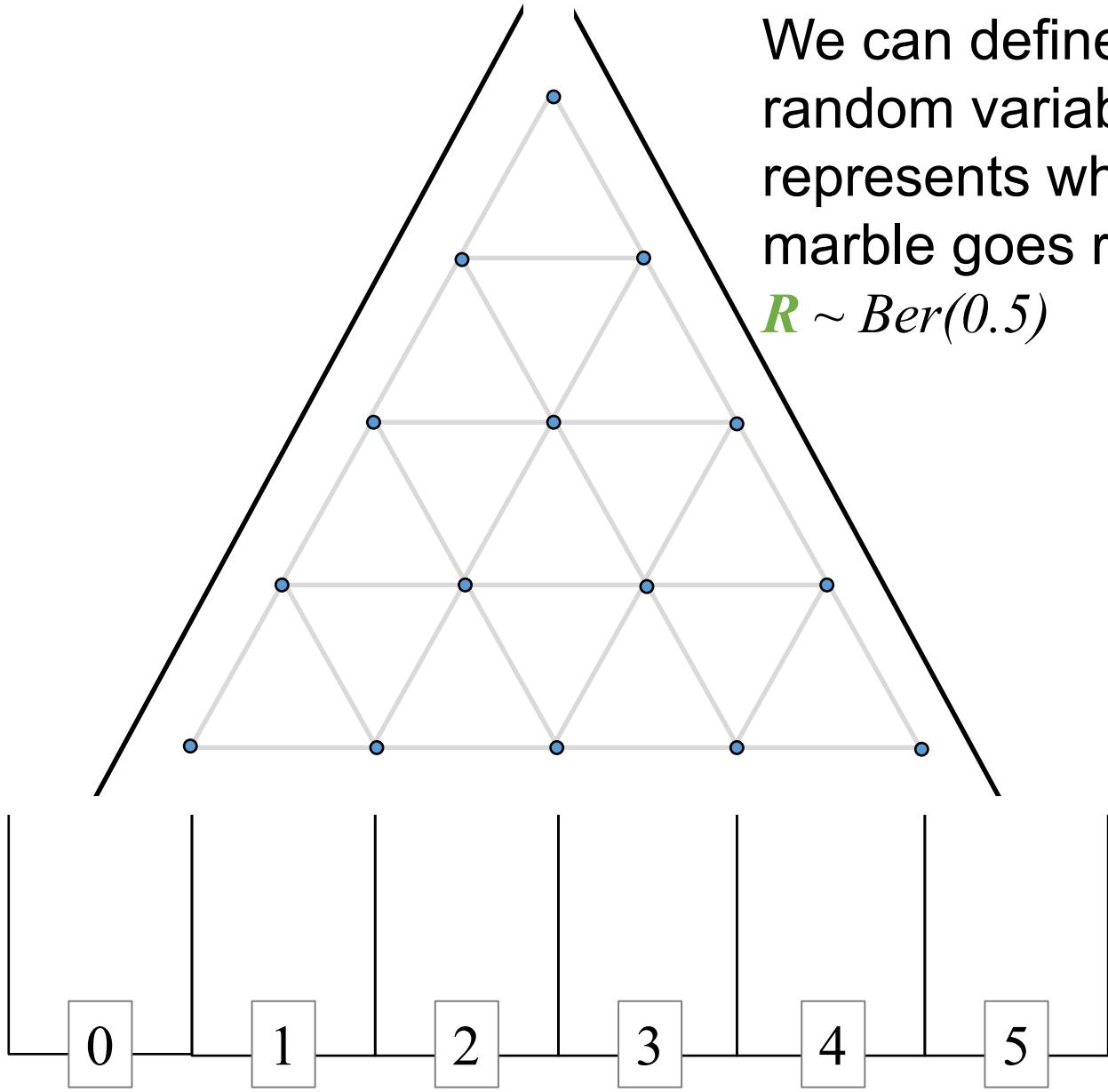
When a marble hits a pin, it has equal chance of going left or right. Each pin represents an independent event.

Galton Board

The bucket index that a marble lands in is equal to the number of times the marble went right



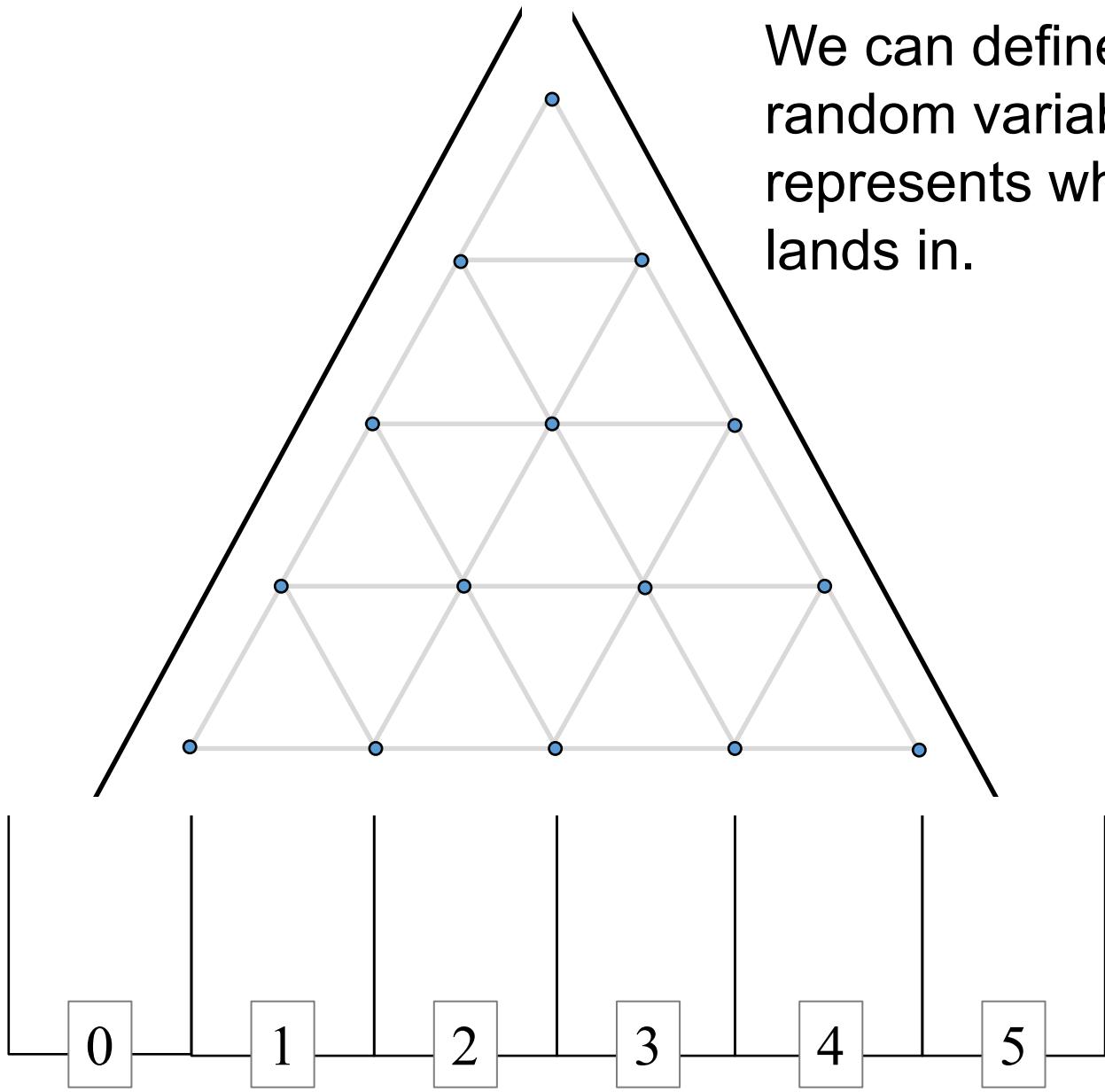
Galton Board



We can define an indicator random variable (R) which represents whether a particular marble goes right as a Bernoulli

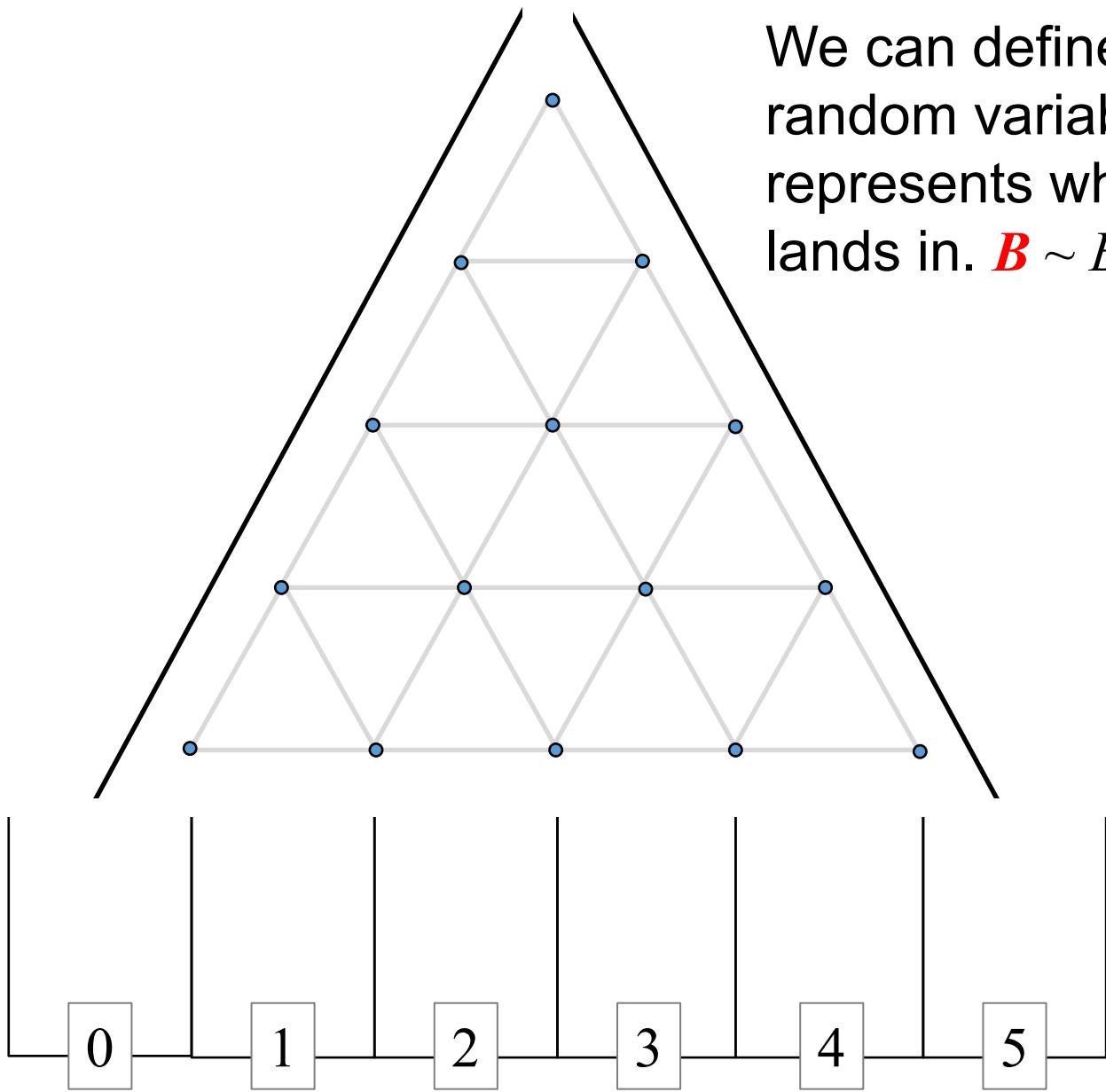
$$R \sim Ber(0.5)$$

Galton Board



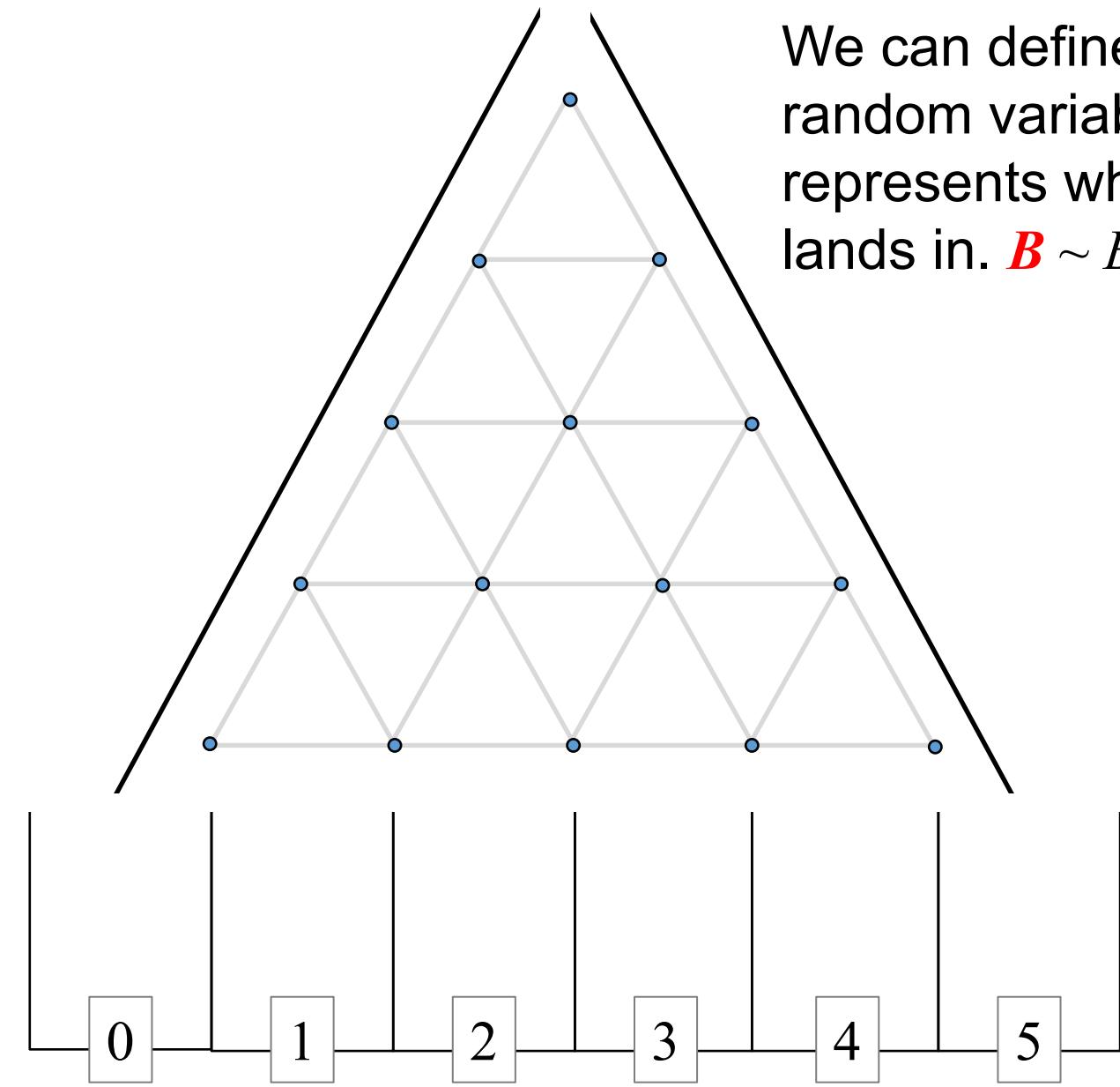
We can define an indicator random variable (B) which represents what bucket a marble lands in.

Galton Board



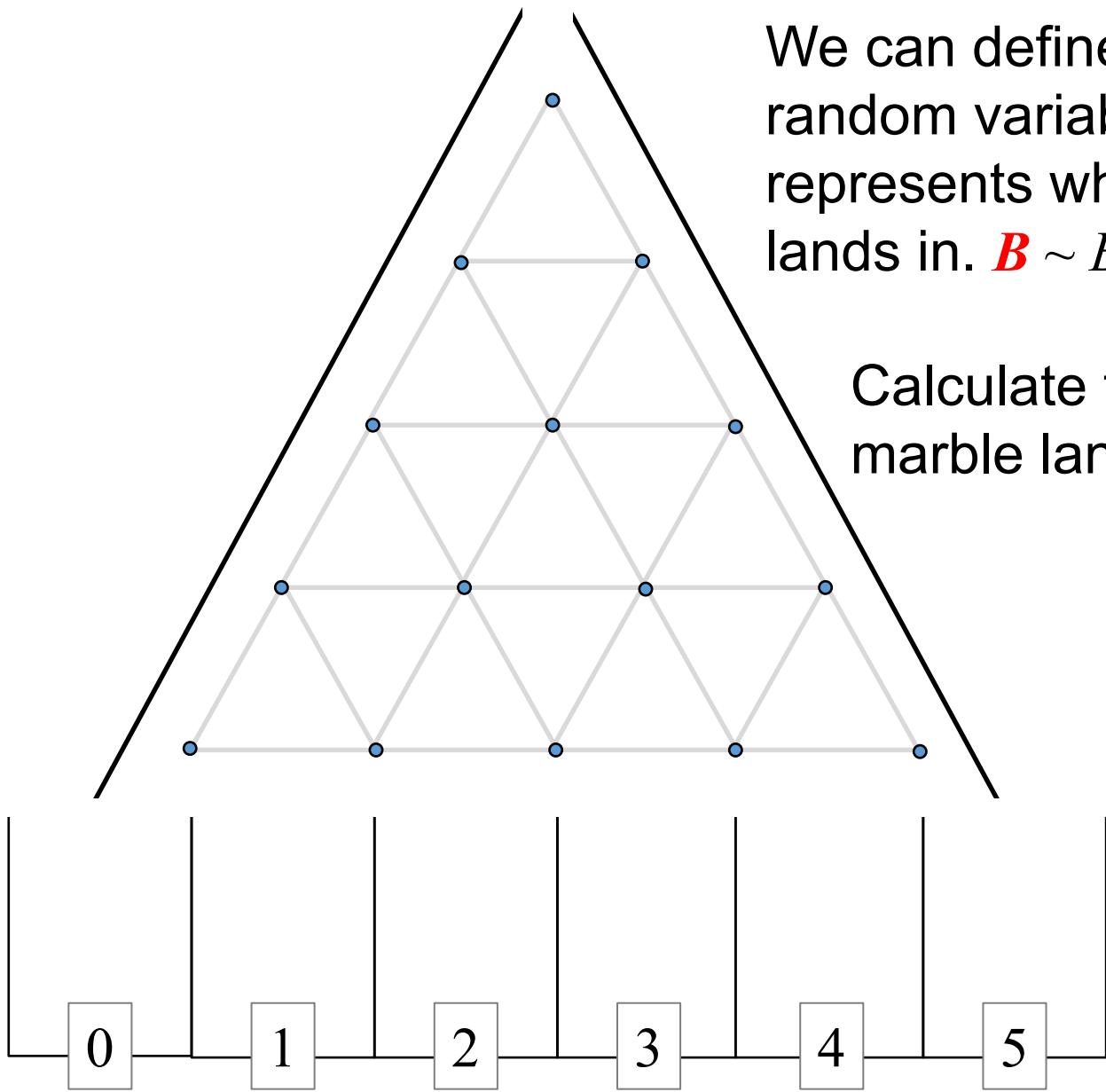
We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim \text{Bin}(\text{levels}, 0.5)$

Galton Board



We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim Bin(5, 0.5)$

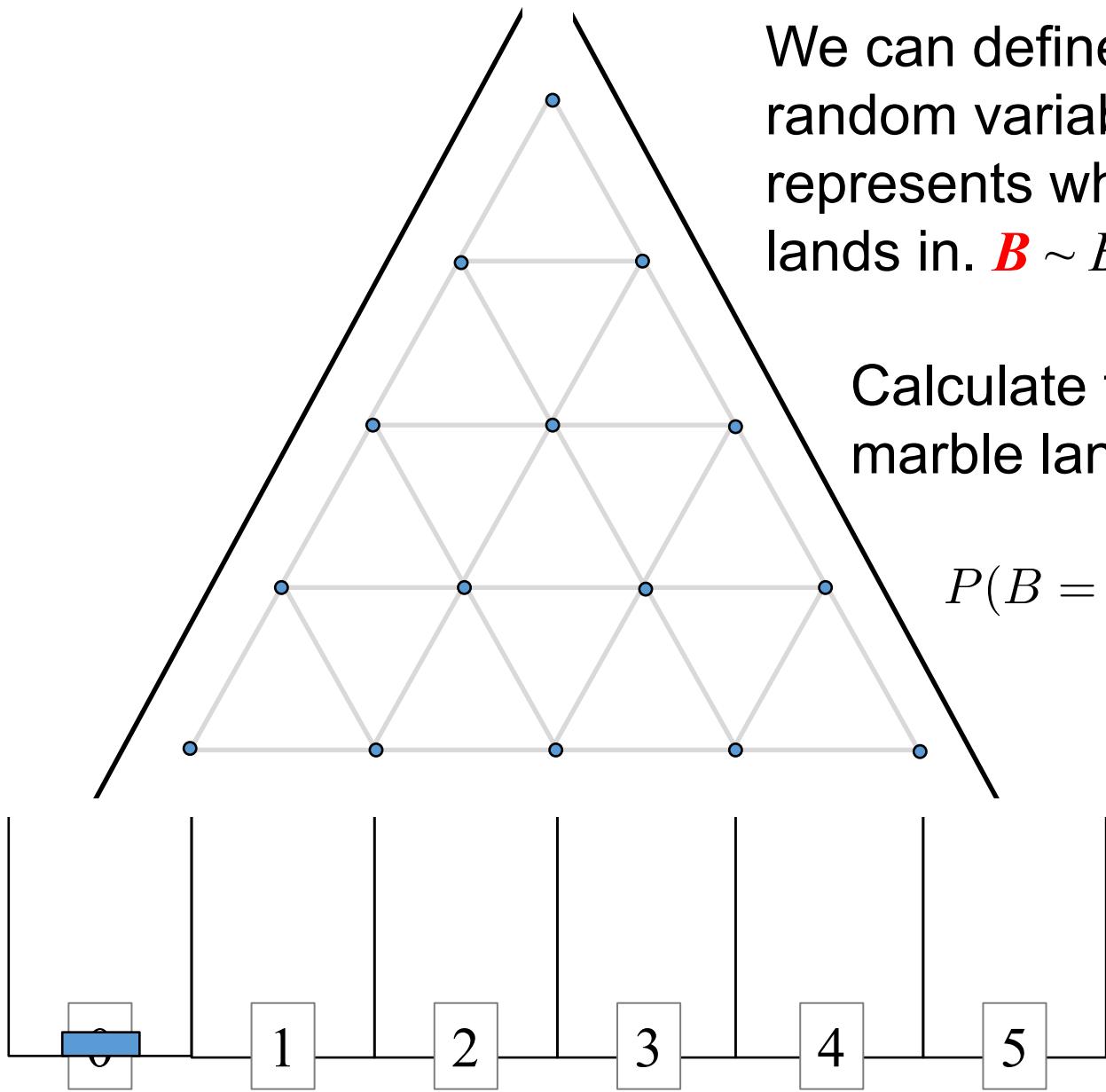
Galton Board



We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim Bin(5, 0.5)$

Calculate the probability of a marble landing in a bucket.

Galton Board

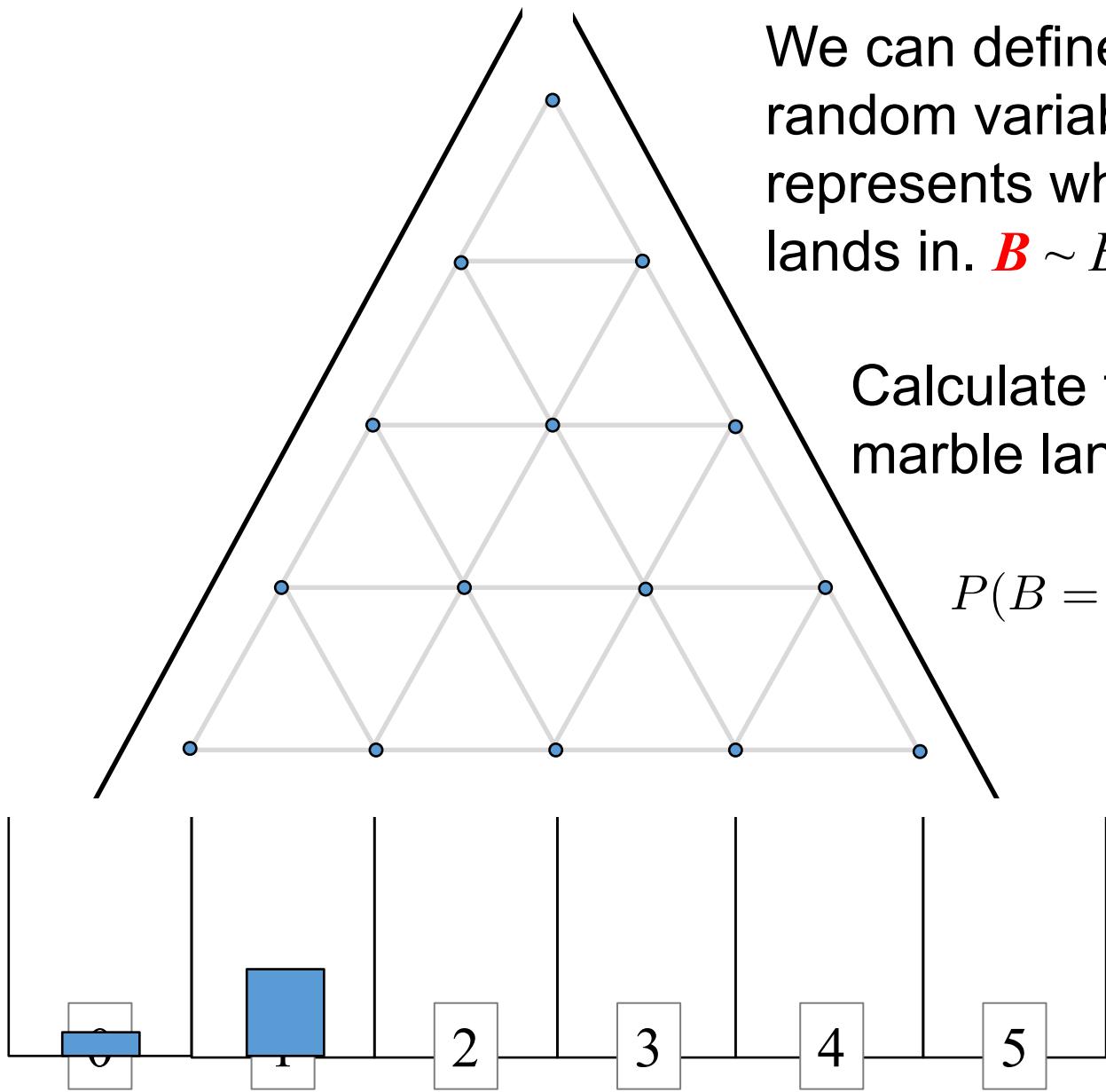


We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim Bin(5, 0.5)$

Calculate the probability of a marble landing in a bucket.

$$P(B = 0) = \binom{5}{0} \frac{1}{2}^5 \approx 0.03$$

Galton Board

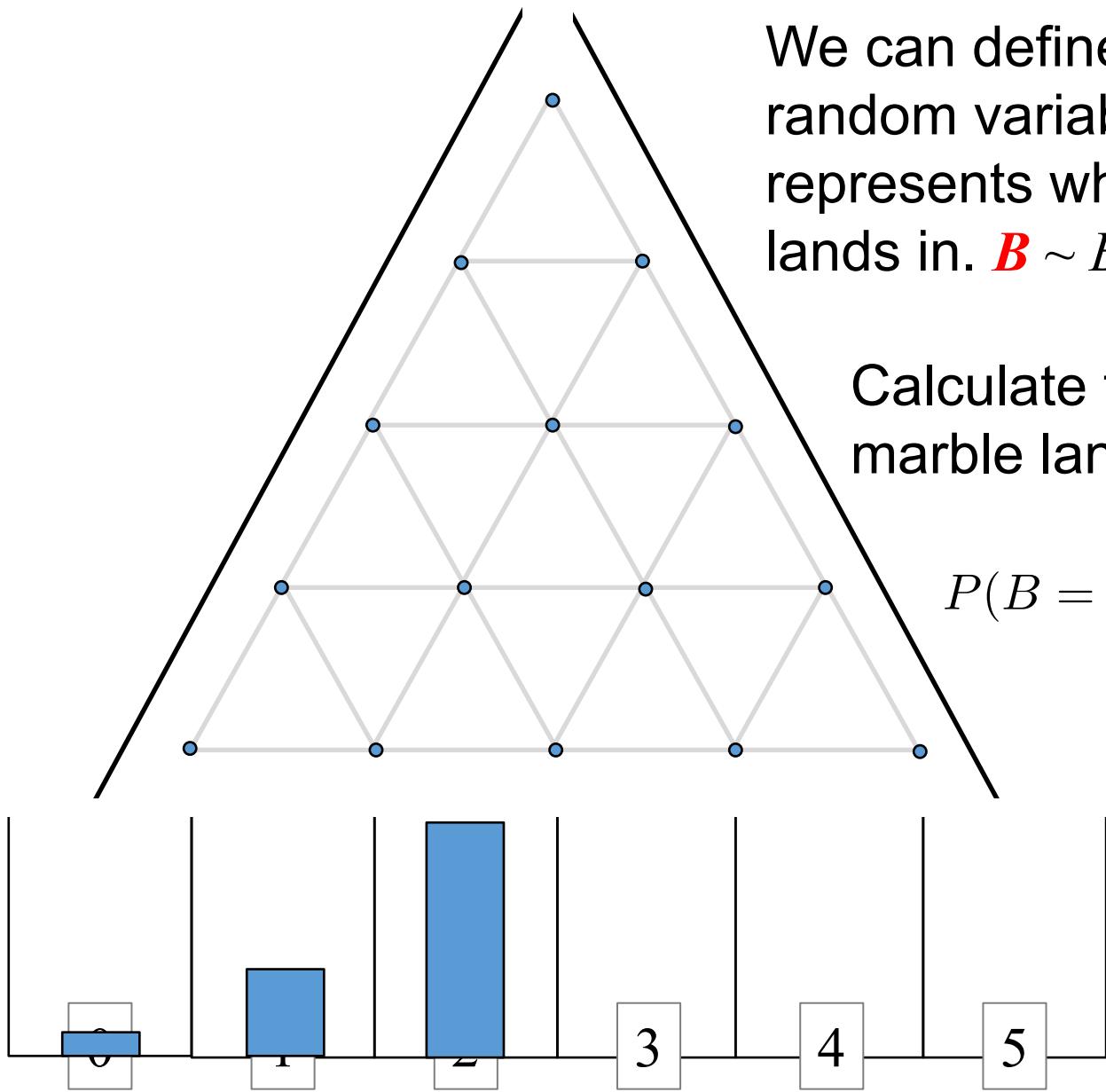


We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim Bin(5, 0.5)$

Calculate the probability of a marble landing in a bucket.

$$P(B = 1) = \binom{5}{1} \frac{1}{2}^5 \approx 0.16$$

Galton Board

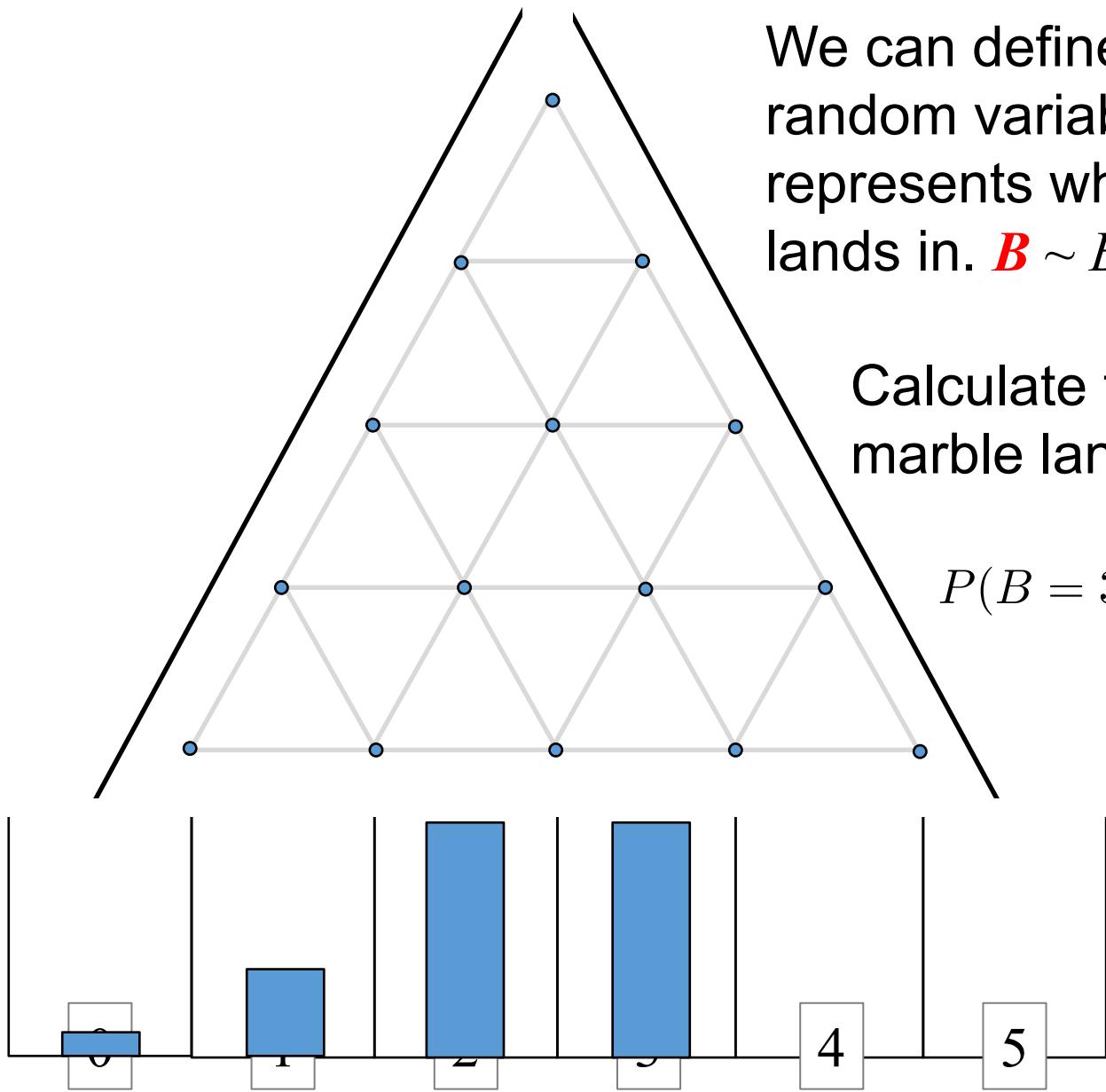


We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim Bin(5, 0.5)$

Calculate the probability of a marble landing in a bucket.

$$P(B = 2) = \binom{5}{2} \frac{1}{2}^5 \approx 0.31$$

Galton Board

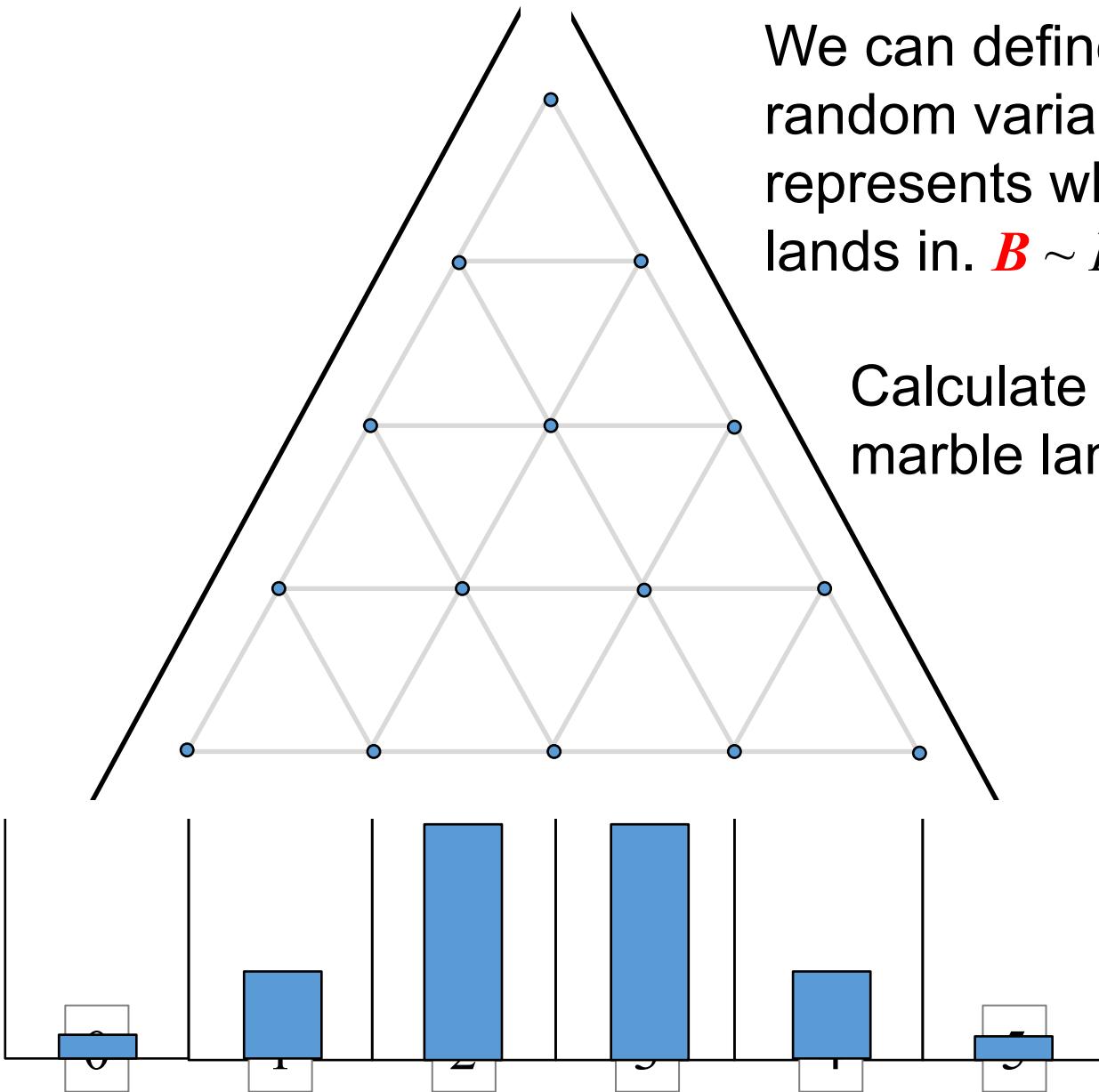


We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim Bin(5, 0.5)$

Calculate the probability of a marble landing in a bucket.

$$P(B = 3) = \binom{5}{2} \frac{1}{2}^5 \approx 0.31$$

Galton Board



We can define an indicator random variable (B) which represents what bucket a marble lands in. $B \sim Bin(5, 0.5)$

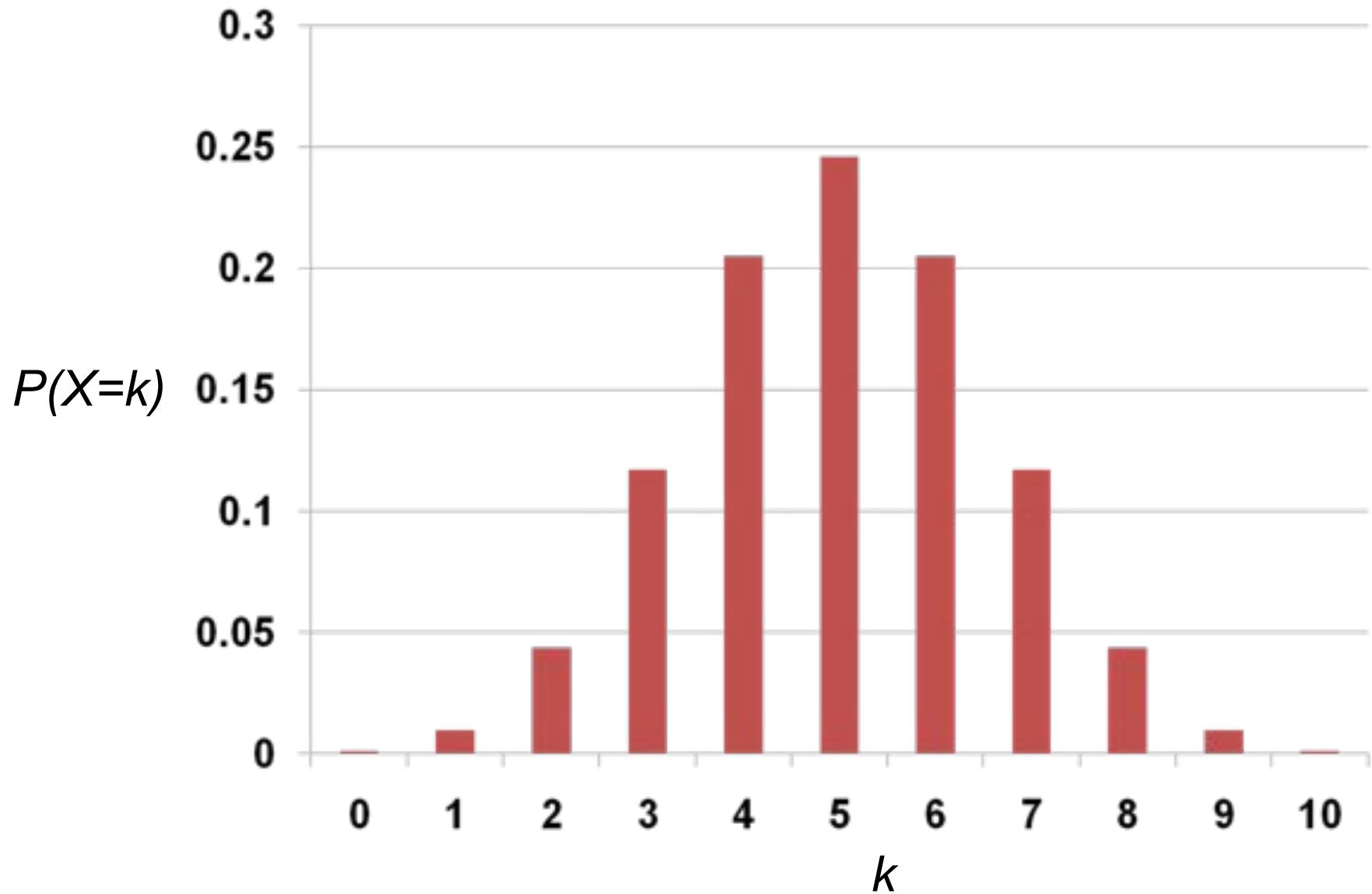
Calculate the probability of a marble landing in a bucket.

PDF

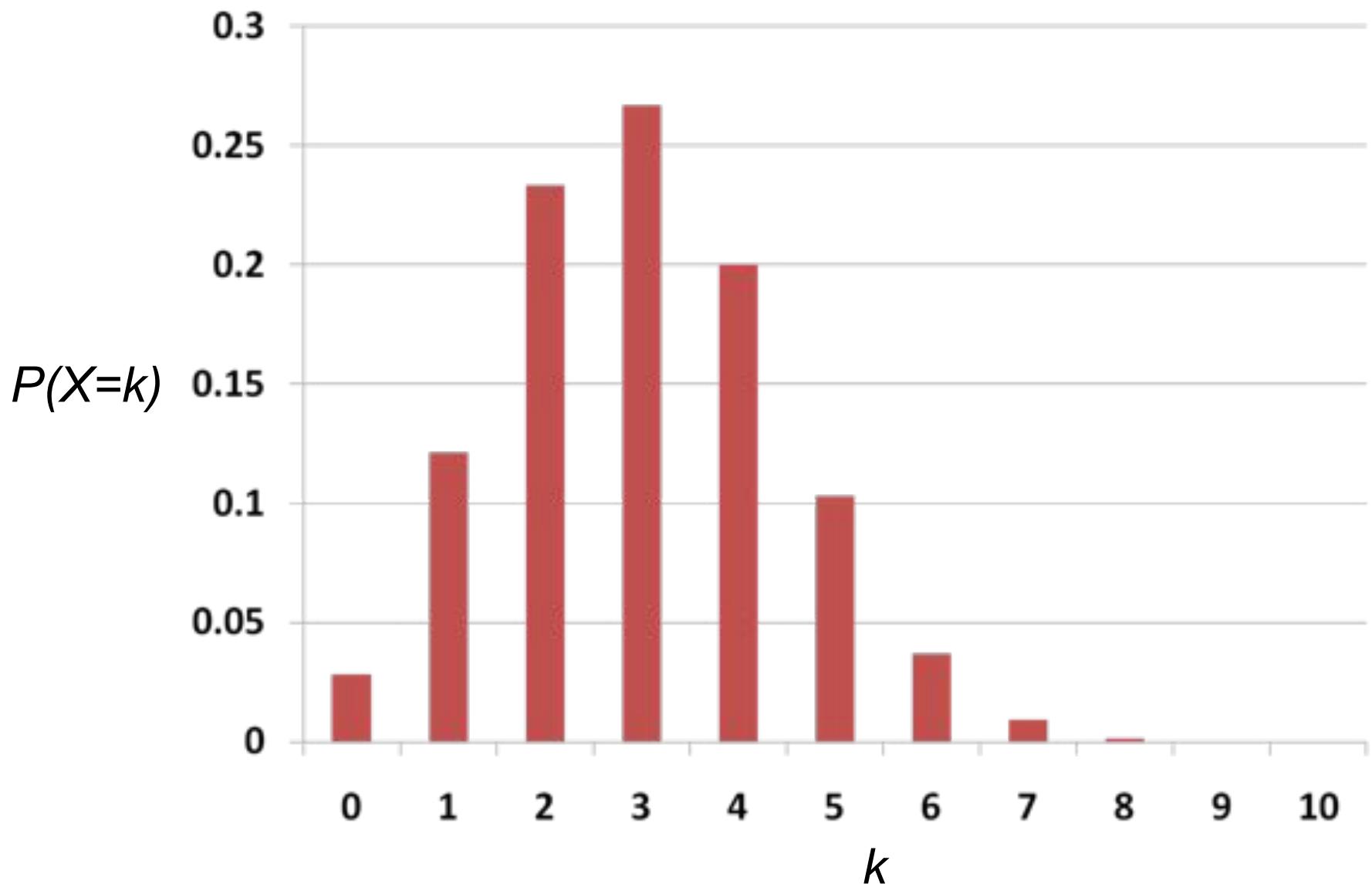


FROM CHAOS TO ORDER

PMF for $X \sim \text{Bin}(n = 10, p = 0.5)$



PMF for $X \sim \text{Bin}(n = 10, p = 0.3)$



Genetic Inheritance

- Person has 2 genes for trait (eye color)
 - Child receives 1 gene (equally likely) from each parent
 - Child has brown eyes if either (or both) genes brown
 - Child only has blue eyes if both genes blue
 - Brown is “dominant” (d) , Blue is “recessive” (r)
 - Parents each have 1 brown and 1 blue gene
- 4 children, what is $P(3 \text{ children with brown eyes})$?
 - Child has blue eyes: $p = (\frac{1}{2})(\frac{1}{2}) = \frac{1}{4}$ (2 blue genes)
 - $P(\text{child has brown eyes}) = 1 - (\frac{1}{4}) = 0.75$
 - $X = \# \text{ of children with brown eyes. } X \sim \text{Bin}(4, 0.75)$

$$P(X = 3) = \binom{4}{3} (0.75)^3 (0.25)^1 \approx 0.4219$$

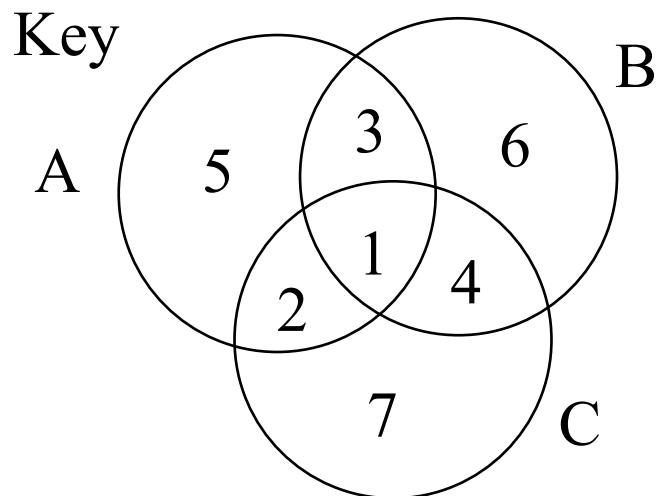
A small, clear plastic model sailboat with yellow and white markings on its hull is positioned in the bottom right corner of the frame. The boat is oriented diagonally, facing towards the top left. The background is a dark, textured surface that appears to be a starry night sky or a dark ocean, with numerous small white dots representing stars or distant lights.

1001

Have original 4 bit string to send over network.

Add 3 “parity” bits and send 7 bits total

Each bit independently corrupted (flipped) in transmission with probability 0.1. What is the probability of successful transmission?



Receive 1110000?

Send 1110?

Receive 1010100?

Have original 4 bit string to send over network.

Add 3 “parity” bits and send 7 bits total

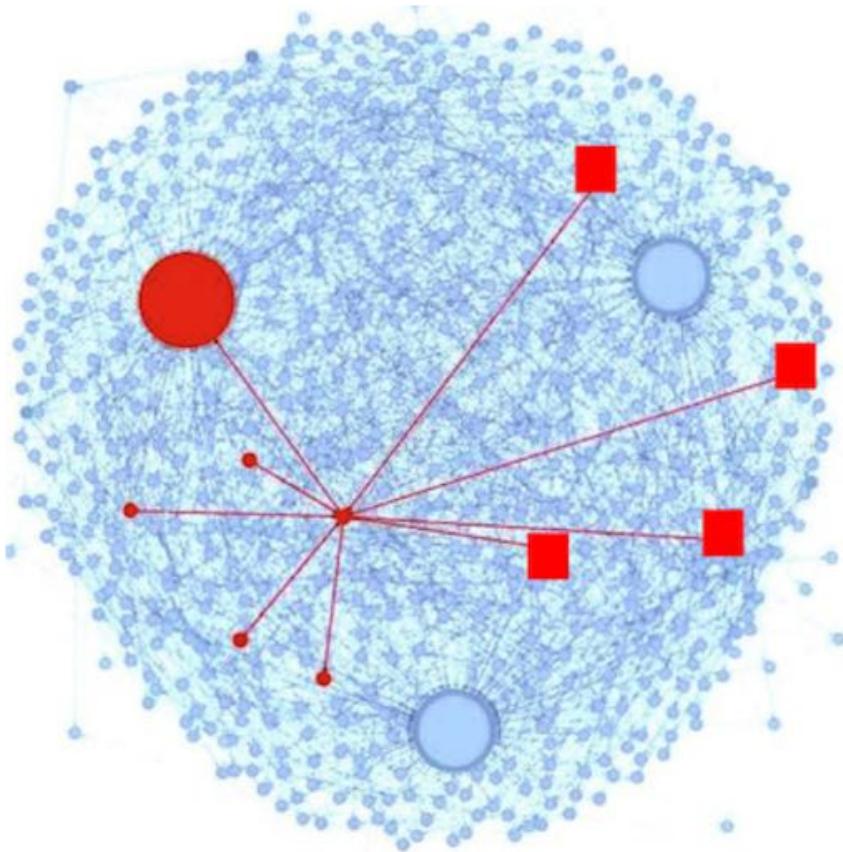
Each bit independently corrupted (flipped) in transmission with probability 0.1. What is the probability of successful transmission?

Three Graders

Three peer graders (A, B, C) grade the same submission for a problem with 100 points. Each grader gives a grade which is a Binomial with $n = 100$, $p = 0.8$. What is the Expected average of their three grades?

Is Peer Grading Accurate Enough?

Looking ahead

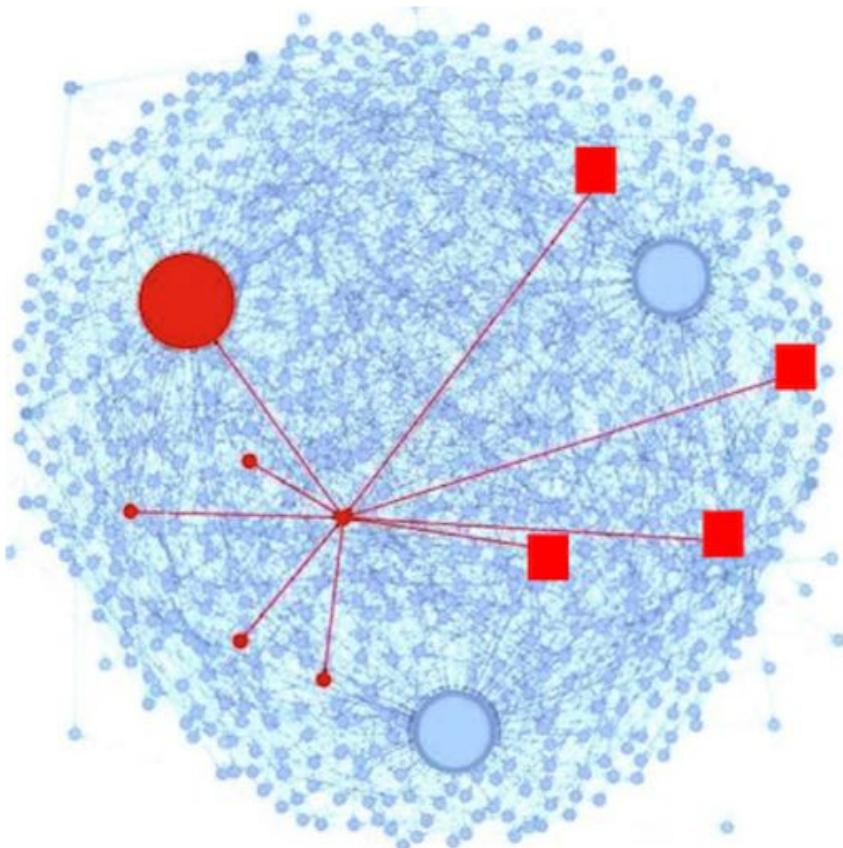


Peer Grading on Coursera
HCI.

31,067 peer grades for
3,607 students.

Is Peer Grading Accurate Enough?

Looking ahead



1. Defined random variables for:
 - True grade (s_i) for assignment i
 - Observed (z_i^j) score for assign i
 - Bias (b_j) for each grader j
 - Variance (r_j) for each grader j
2. Designed a probabilistic model that defined the distributions for all random variables

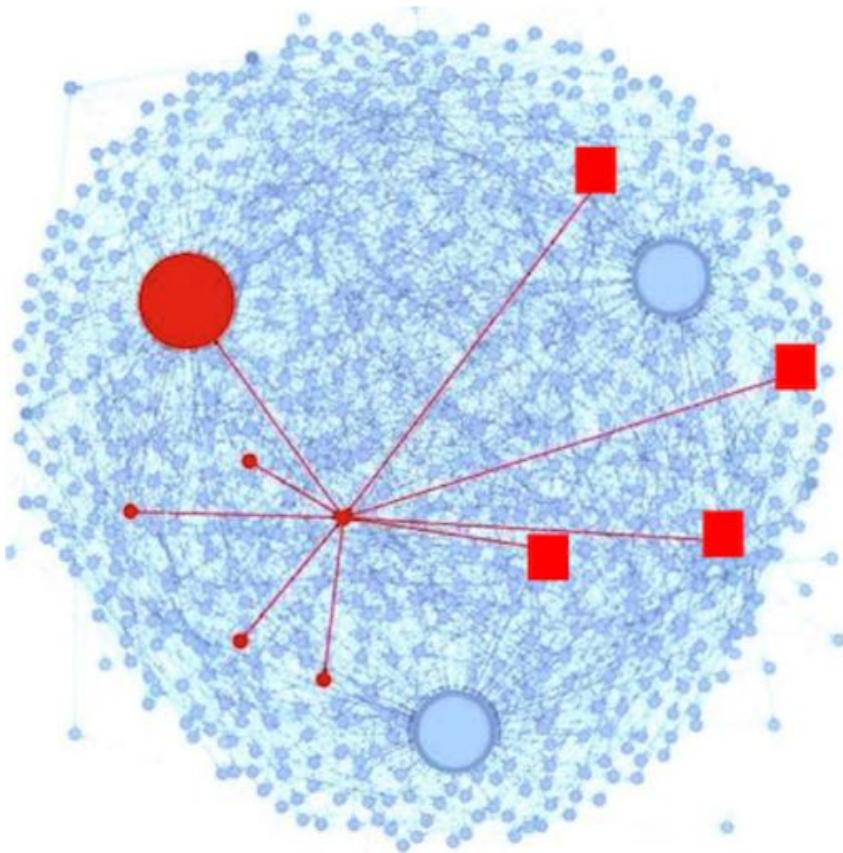
$$s_i \sim \text{Bin}(\text{points}, \theta)$$

Problem param

$$z_i^j \sim \mathcal{N}(\mu = s_i + b_j, \sigma = \sqrt{r_j})$$

Is Peer Grading Accurate Enough?

Looking ahead

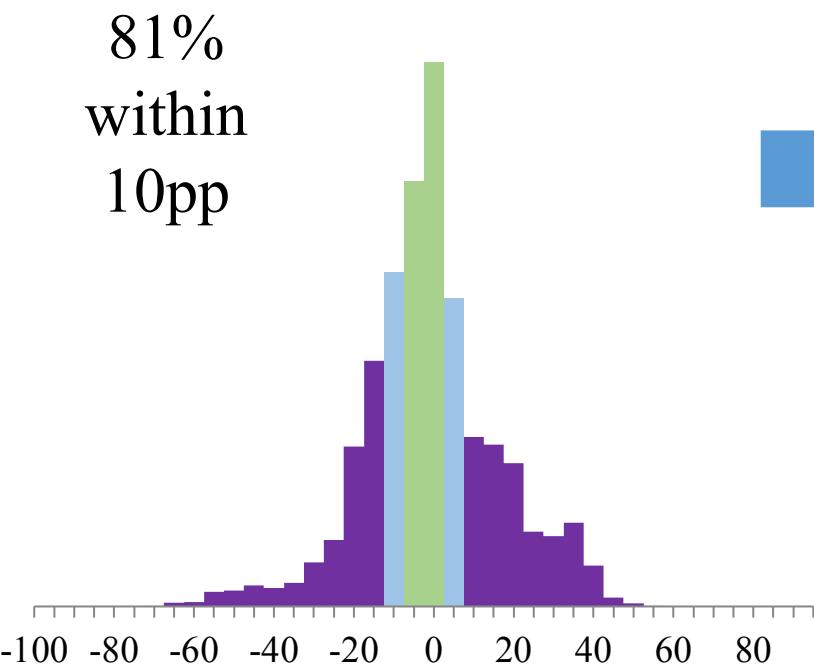


1. Defined random variables for:
 - True grade (s_i) for assignment i
 - Observed (z_i^j) score for assign i
 - Bias (b_j) for each grader j
 - Variance (r_j) for each grader j
2. Designed a probabilistic model that defined the distributions for all random variables
3. Found the variable assignments that maximized the probability of our observed data

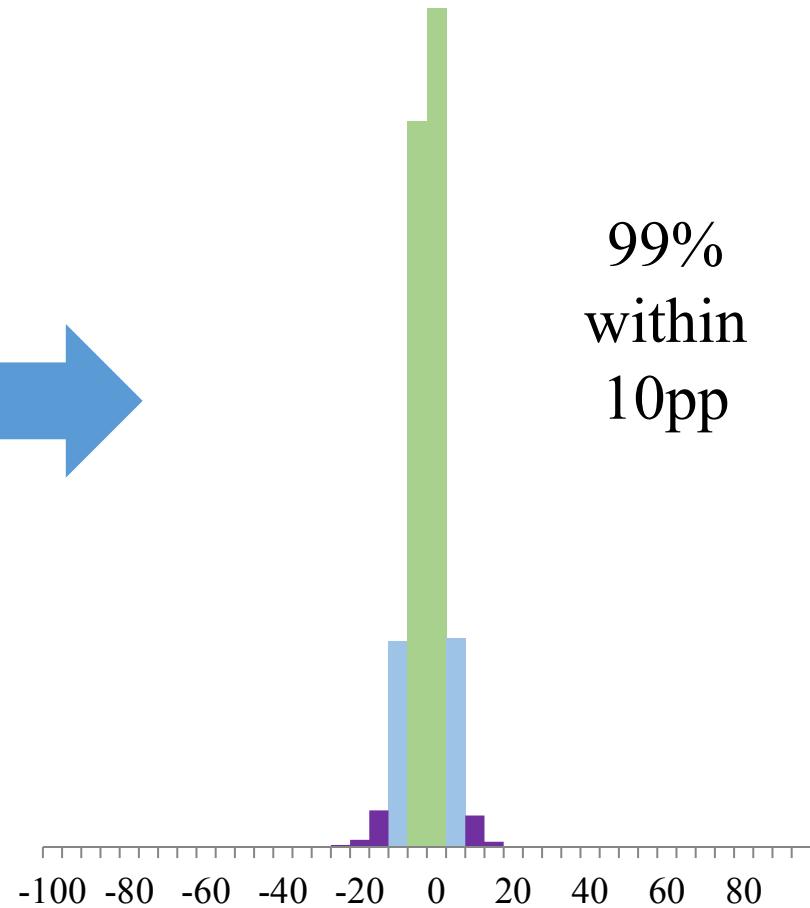
Inference or Machine Learning

Yes, With Probabilistic Modelling

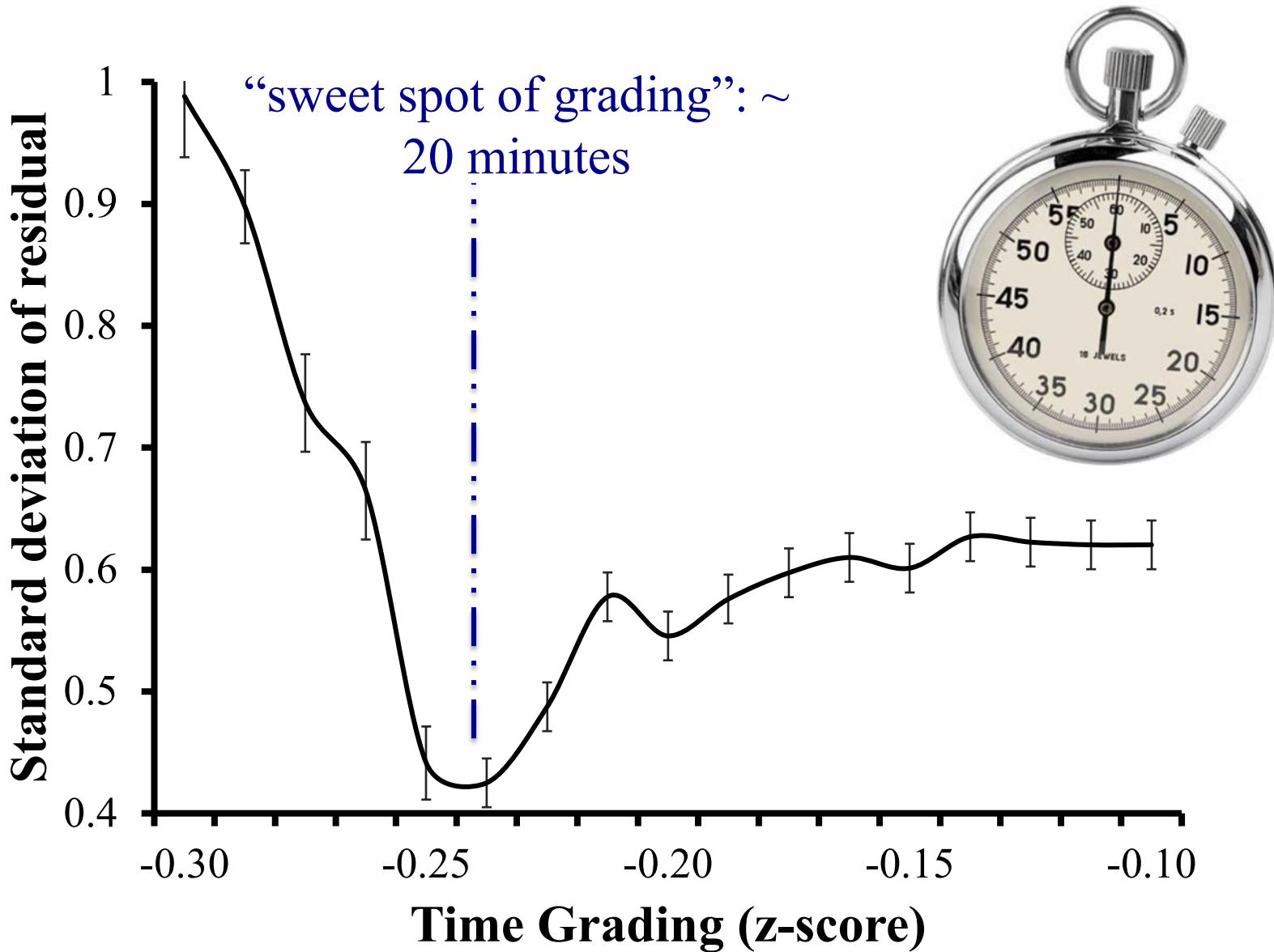
Before:



After:



Grading Sweet Spot



Voilà, c'est tout

