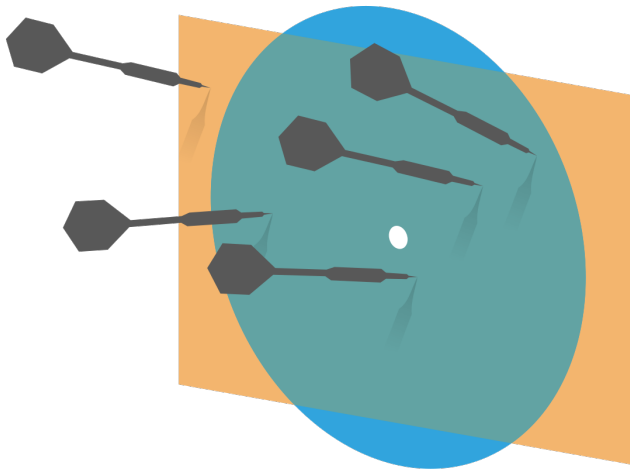


Probability and Statistics for Computer Science



Credit: wikipedia

Who discovered this?

$$e = \lim_{n \rightarrow \infty} \left(1 + \frac{1}{n} \right)^n$$

Last time

- ✱ Random Variable

- ✱ *Review , Covariance*

- ✱ *The weak law of large numbers*

Proof of Weak law of large numbers

- * Apply Chebyshev's inequality

$\forall \epsilon > 0$

$$P(|\bar{\mathbf{X}} - E[\bar{\mathbf{X}}]| \geq \epsilon) \leq \frac{\text{var}[\bar{\mathbf{X}}]}{\epsilon^2}$$

- * Substitute $E[\bar{\mathbf{X}}] = E[X]$ and $\text{var}[\bar{\mathbf{X}}] = \frac{\text{var}[X]}{N}$

$$P(|\bar{\mathbf{X}} - E[X]| \geq \epsilon) \leq \frac{\text{var}[X]}{N\epsilon^2} \xrightarrow{N \rightarrow \infty} 0$$

$$\lim_{N \rightarrow \infty} P(|\bar{\mathbf{X}} - E[X]| \geq \epsilon) = 0$$

Random Variable Example

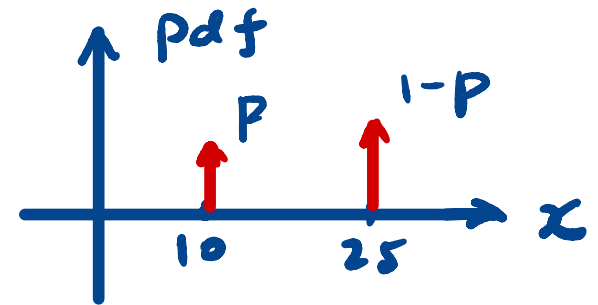


Money box

* shake and take one and put back.

X_1, X_2, \dots, X_N
are iid.

10¢ 25¢
dime quarter



X_1 takes $x_1=10$

X_2 takes $x_2=10$

X_3 takes $x_3=25$

\vdots

X_N takes $x_N=10$

$$E[X] = ? \cdot 10P + (1-P) \cdot 25$$

$$\bar{X} = \frac{\sum X_i}{N}$$

Applications of the Weak law of large numbers

- ✱ The law of large numbers *justifies using simulations* (instead of calculation) to estimate the expected values of random variables HW4

$$\lim_{N \rightarrow \infty} P(|\bar{X} - E[X]| \geq \epsilon) = 0$$

- ✱ The law of large numbers also *justifies using histogram* of large random samples to approximate the probability distribution function $P(x)$, see proof on Pg. 353 of the textbook by DeGroot, et al.

Histogram of large random IID samples approximates the probability distribution

✱ The law of large numbers justifies using histograms to approximate the probability distribution. Given N IID random variables X_1, \dots, X_N

✱ According to the law of large numbers

$$\bar{Y} = \frac{\sum_{i=1}^N Y_i}{N} \xrightarrow{N \rightarrow \infty} E[Y_i]$$

✱ As we know for indicator function *textbook pg 100-101*

$$E[Y_i] = P(c_1 \leq X_i < c_2) = P(c_1 \leq X < c_2)$$

What is the number?

$$(p+q)^n = \sum_{k=0}^n a_k p^k q^{n-k}$$

$$a_k = ? \quad \binom{n}{k} \rightarrow \text{Binomial Coeff.}$$

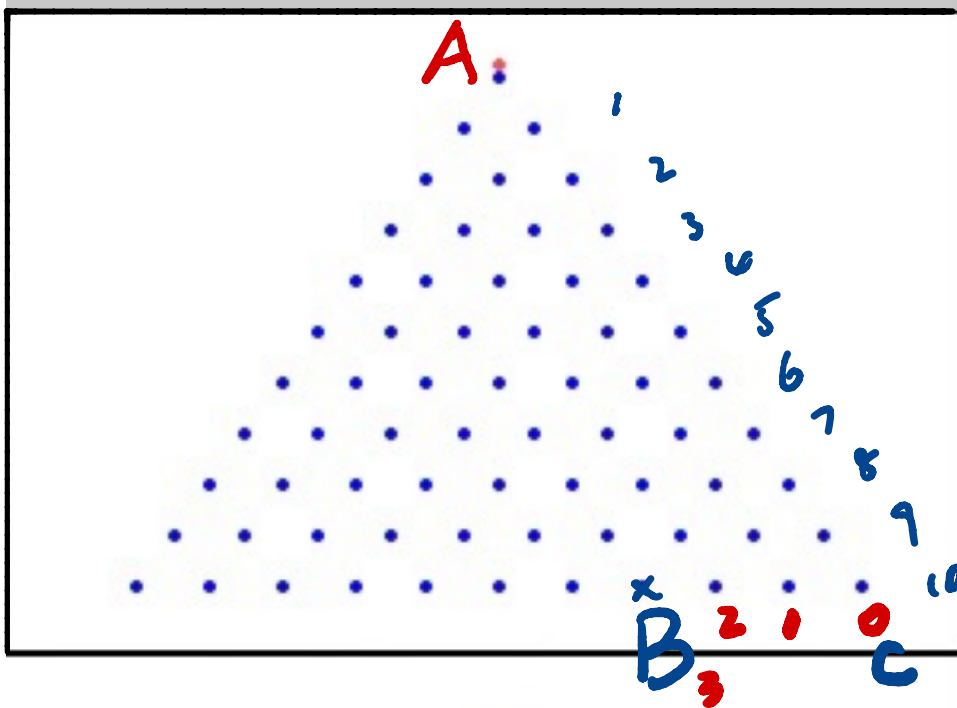
if $p+q=1$

$$\text{LHS} = 1^n = 1 \quad \text{RH} = 1$$

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

How many paths that lead $A \rightarrow B$?

$$\frac{L}{R?} - - - \frac{L}{R?} - - - - - \frac{L}{10}$$



$$\binom{10}{3} \text{ or } \binom{10}{7}$$

$$\binom{10}{k}$$

Objectives

- ✱ Important known discrete probability distributions
- ✱ Continuous Random Variable

The classic discrete distributions

✱ Bernoulli

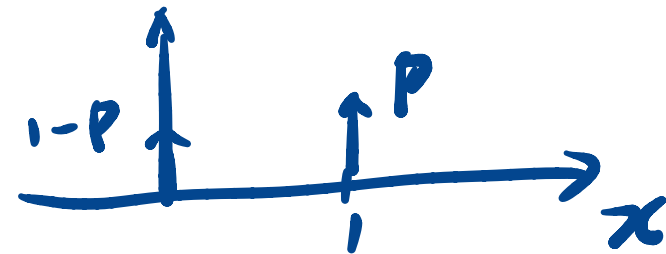
H T coin

✱ Binomial *sum of N*

Bernoulli:

$$X(\omega) = \begin{cases} 1 & H \\ 0 & H^c \end{cases}$$

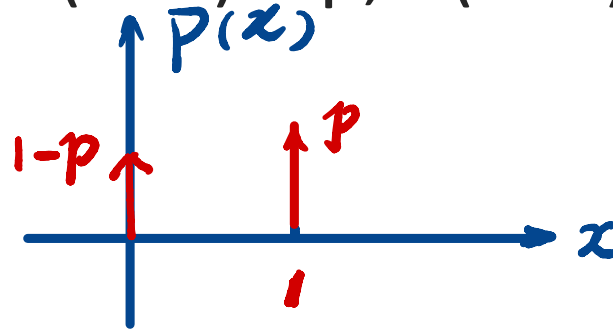
✱ Geometric



✱ Discrete uniform

Bernoulli distribution

- ✱ A random variable X is **Bernoulli** if it takes on two values 0 and 1 such that $P(X=1) = p$, $P(X=0)=1-p$



$$E[X] = p$$

$$\text{var}[X] = p(1 - p)$$

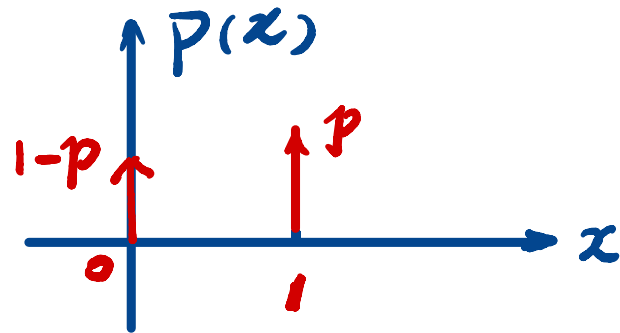


Jacob Bernoulli (1654-1705)

Credit: wikipedia

Bernoulli distribution

$$\begin{aligned} E[X] &= p \cdot 1 + (1-p) \cdot 0 \\ &= p \end{aligned}$$



$$\begin{aligned} \text{var}[X] &= E[X^2] - (E[X])^2 \\ &= 0^2 \cdot (1-p) + 1^2 \cdot p - p^2 \\ &= p - p^2 \\ &= p(1-p) \end{aligned}$$

Bernoulli distribution

* Examples

- * Tossing a biased (or fair) coin
- * Making a free throw
- * Rolling a six-sided die and checking if it shows 6
- * **Any indicator function** of a random variable

$$I_A = \begin{cases} 1 & \text{event } A \text{ happens} \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} E[I_A] &= 1 \cdot P(A) + 0 \cdot \cancel{P(A^c)} \\ &= P(A) \end{aligned}$$

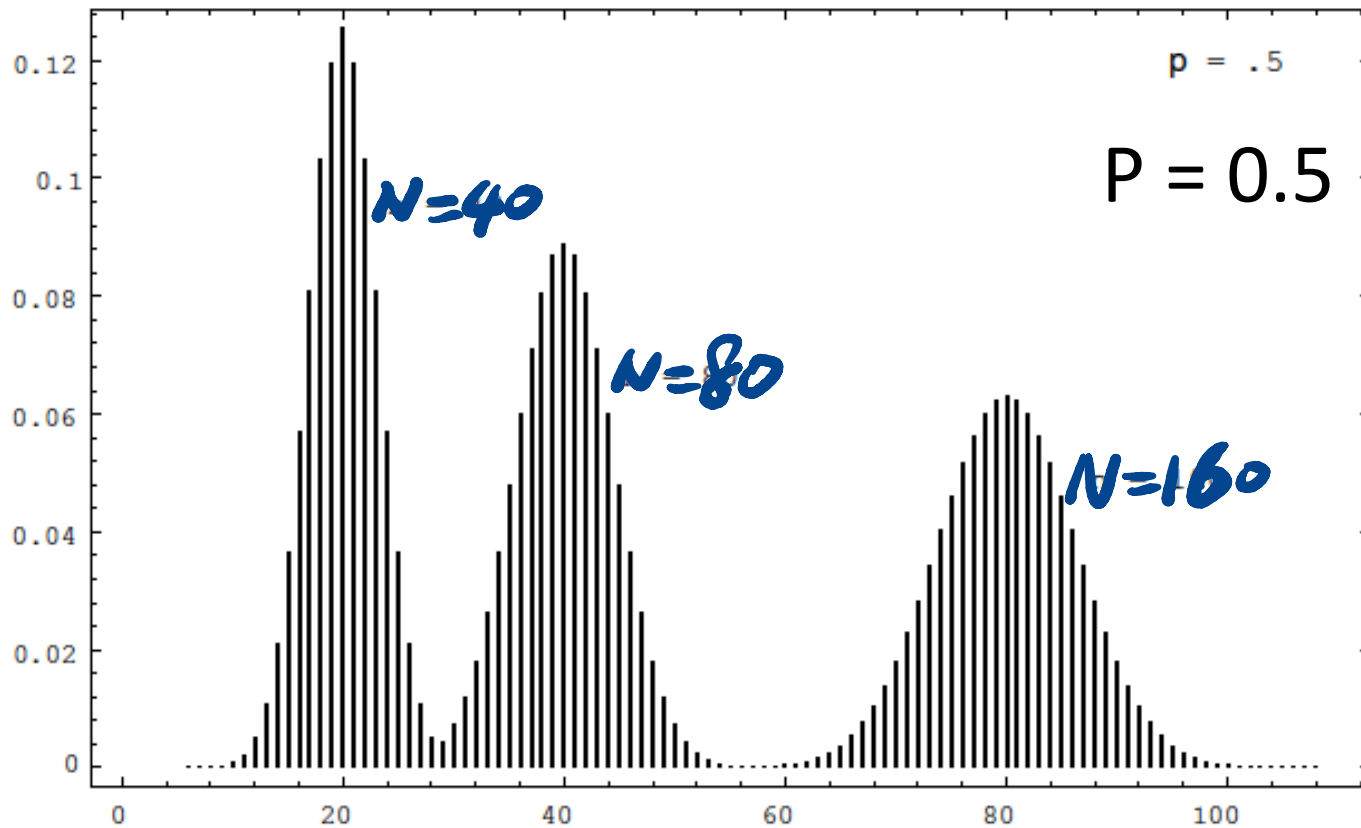
Binomial distribution

- ✱ The Galton Board

[http://www.randomservices.org/
random/apps/
GaltonBoardExperiment.html](http://www.randomservices.org/random/apps/GaltonBoardExperiment.html)

- ✱ Remember the airline problem?

Binomial distribution



N

p

Credit: Prof. Grinstead

Binomial distribution

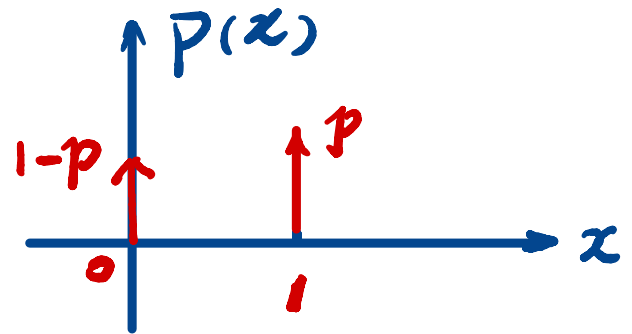
Binomial RV X_S is the sum of N independent Bernoulli RVs

$$X_S = \sum_{i=1}^N X_i$$

$$X_i(\omega) = \begin{cases} 1 & \omega \text{ is event } A \\ 0 & \omega \text{ is } A^c \end{cases}$$

Range of X_S is?
 $[0, N]$

Number of heads
in N tosses.



Binomial distribution



$$N=7 \quad P(X_S=3)=?$$

$$P(X_S=3) = \sum P(\text{any one instance of } 3Hs)$$

Consider random tossing 7 times a biased coin

$$P(X_S=3) = \binom{7}{3} p^3 (1-p)^4$$

$$\binom{7}{3} P(c_1=H, c_2=T, c_3=H, \dots) \\ p(1-p) \cdot p(1-p)(1-p) \cdot p(1-p) \dots \\ p^3 (1-p)^4$$

Binomial distribution

N 0 0 0 $-$ $-$ $-$ 0

K^{th}
heads

$$P(X=K) = \binom{N}{K} p^K (1-p)^{N-K}$$

$$N \geq K \geq 0$$

$$(p+q)^n = \sum \binom{n}{k} p^k q^{n-k} = 1$$

if $q=1-p$

Binomial distribution

- ✱ A discrete random variable X is binomial if

$$P(X = k) = \binom{N}{k} p^k (1 - p)^{N-k} \quad \text{for integer } 0 \leq k \leq N$$

with $E[X] = Np$ & $var[X] = Np(1 - p)$

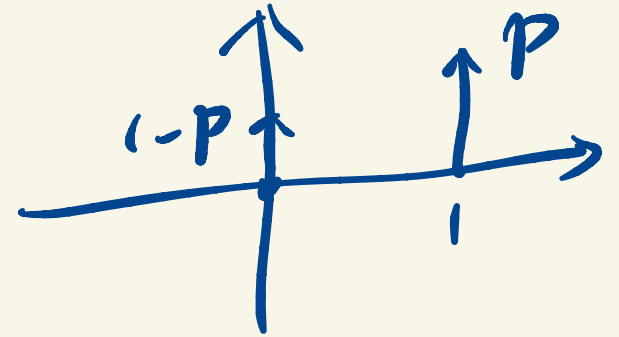
- ✱ Examples

- ✱ If we roll a six-sided die N times, how many sixes we will see
- ✱ If I attempt N free throws, how many points will I score
- ✱ **What is the sum of N independent and identically distributed Bernoulli trials?**

$$X_S = \sum_{i=1}^N X_i$$

$$X_i = \begin{cases} 1 & \text{H} \\ 0 & \text{T} \end{cases}$$

$$\begin{aligned} E[X_S] &= \sum E[X_i] \\ &= N \cdot p \end{aligned}$$



$$E[X_i] = p$$

$$\text{var}[X_S] = \text{var}[\sum X_i]$$

$\therefore X_i$ are indpt

X_i are
iid

$$= \sum (\text{var}[X_i])_{\text{iid}}$$

$$\text{var}[X_i] = p(1-p)$$


$$= N \cdot p(1-p)$$

Expectations of Binomial distribution

✱ A discrete random variable X is binomial if

$$P(X = k) = \binom{N}{k} p^k (1 - p)^{N-k} \quad \text{for integer } 0 \leq k \leq N$$

with $E[X] = Np$ & $var[X] = Np(1 - p)$



Binomial distribution: die example

- Let X be the number of sixes in 36 rolls of a fair six-sided die. What is $P(X=k)$ for $k=5, 6, 7$

$$P(X=6) = \binom{36}{6} \left(\frac{1}{6}\right)^6 \left(\frac{5}{6}\right)^{36-6}$$

$$P(X \geq 6) = \sum_{k=6}^{36} \binom{36}{k} \left(\frac{1}{6}\right)^k \left(\frac{5}{6}\right)^{36-k}$$

- Calculate $E[X]$ and $\text{var}[X]$

$$1 - \sum_{k=0}^5 \dots$$

$$E[X] = 36 \times \frac{1}{6} = 6 \quad \text{var}[X] = 36 \times \frac{1}{6} \cdot \frac{5}{6} = 5$$

Geometric distribution

- ✱ A discrete random variable X is geometric if

$$P(X = k) = (1 - p)^{k-1} p \quad k \geq 1$$

H, TH, TTH, TTTH, TTTTH, TTTTTH, ...

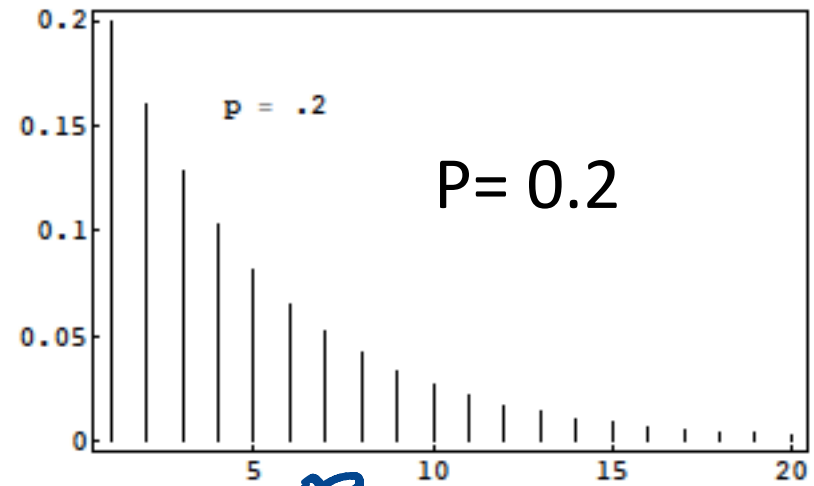
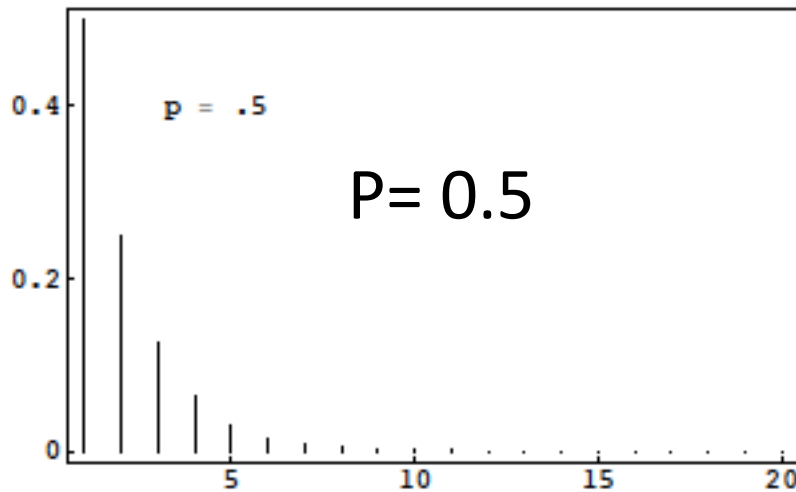
(Note: In the original image, blue brackets are drawn under the 'TTTTTH' sequence. One bracket above the 'T's' is labeled 'k-1', and one bracket below the entire sequence is labeled 'k'.)

- ✱ Expected value and variance

$$E[X] = \frac{1}{p} \quad \& \quad \text{var}[X] = \frac{1 - p}{p^2}$$

Geometric distribution

$$P(X = k) = (1 - p)^{k-1} p \quad k \geq 1$$



$$E[X] = \sum_{x=1}^{\infty} x P(x) = \frac{1}{p}$$

Credit: Prof. Grinstead

Geometric distribution

✱ Examples:

- ✱ How many rolls of a six-sided die will it take to see the first 6?
- ✱ How many Bernoulli trials must be done before the first 1?
- ✱ How many experiments needed to have the first success?
- ✱ Plays an important role in the **theory of queues**

Derivation of geometric expected value

$$E[X] = \sum_{k=1}^{\infty} k(1-p)^{k-1}p$$

Derivation of geometric expected value

$$E[X] = \sum_{k=1}^{\infty} k(1-p)^{k-1}p$$

$$= p \sum_{k=1}^{\infty} k(1-p)^{k-1}$$

$\alpha < 1$

$$\sum_{n=1}^{\infty} n \alpha^{n-1} =$$

Derivation of geometric expected value

$$\begin{aligned} E[X] &= \sum_{k=1}^{\infty} k(1-p)^{k-1}p \\ &= p \sum_{k=1}^{\infty} k(1-p)^{k-1} \\ &= \frac{p}{1-p} \sum_{k=1}^{\infty} k(1-p)^k \end{aligned}$$

Derivation of geometric expected value

$$\begin{aligned} E[X] &= \sum_{k=1}^{\infty} k(1-p)^{k-1}p \\ &= p \sum_{k=1}^{\infty} k(1-p)^{k-1} \\ &= \frac{p}{1-p} \sum_{k=1}^{\infty} k(1-p)^k \end{aligned}$$

✱ For we have

this power series:

Derivation of geometric expected value

$$E[X] = \sum_{k=1}^{\infty} k(1-p)^{k-1}p$$

$$= p \sum_{k=1}^{\infty} k(1-p)^{k-1}$$

$$= \frac{p}{1-p} \sum_{k=1}^{\infty} k(1-p)^k$$

✱ For we have

this power series:

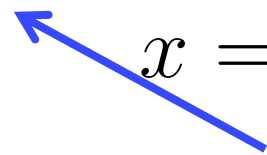
$$\sum_{n=1}^{\infty} nx^n = \frac{x}{(1-x)^2}; \quad |x| < 1$$

Derivation of geometric expected value

$$E[X] = \sum_{k=1}^{\infty} k(1-p)^{k-1}p$$

$$= p \sum_{k=1}^{\infty} k(1-p)^{k-1}$$

$$= \frac{p}{1-p} \sum_{k=1}^{\infty} k(1-p)^k$$

$$x = 1 - p$$


✱ For we have

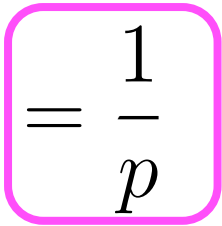
this power series:

$$\sum_{n=1}^{\infty} nx^n = \frac{x}{(1-x)^2}; \quad |x| < 1$$

Derivation of geometric expected value

$$E[X] = \sum_{k=1}^{\infty} k(1-p)^{k-1}p$$

$$= p \sum_{k=1}^{\infty} k(1-p)^{k-1}$$

$$= \frac{p}{1-p} \sum_{k=1}^{\infty} k(1-p)^k$$


✱ For we have


this power series:

$$\sum_{n=1}^{\infty} nx^n = \frac{x}{(1-x)^2}; \quad |x| < 1$$

Derivation of the power series

$$S(x) = \sum_{n=1}^{\infty} nx^n = \frac{x}{(1-x)^2}; \quad |x| < 1$$

Proof: $\frac{S(x)}{x} = \sum_{n=1}^{\infty} nx^{n-1}; \quad \sum_{n=0}^{\infty} x^n = \frac{1}{1-x}; \quad |x| < 1$

$$\int_0^x \frac{S(t)}{t} = \sum_{n=1}^{\infty} x^n = x \cdot \frac{1}{1-x} = \frac{x}{1-x}$$


$$\frac{S(x)}{x} = \left(\frac{x}{1-x} \right)'$$

$$S(x) = \frac{x}{(1-x)^2}$$

Geometric distribution: die example

- Let X be the number of rolls of a fair six-sided die needed to see the first 6. What is $P(X = k)$ for $k = 1, 2$?

$$P(X=1) = \frac{1}{6}$$

$$P(X=2) = \frac{5}{6} \cdot \frac{1}{6}$$

T H with $p = \frac{1}{6}$

- Calculate $E[X]$ and $\text{var}[X]$

$$E[X] = \frac{1}{p} \quad \& \quad \text{var}[X] = \frac{1-p}{p^2}$$

Betting brainteaser

- ✱ What would you rather bet on?
 - ✱ How many rolls of a fair six-sided die will it take to see the first 6?
 - ✱ How many sixes will appear in 36 rolls of a fair six-sided die?

- ✱ Why?

Multinomial distribution

- ✱ A discrete random variable X is Multinomial if

$$P(X_1 = n_1, X_2 = n_2, \dots, X_k = n_k) = \frac{N!}{n_1! n_2! \dots n_k!} p_1^{n_1} p_2^{n_2} \dots p_k^{n_k}$$

where $N = n_1 + n_2 + \dots + n_k$

- ✱ The event of throwing N times the k -sided die to see the probability of getting $n_1 X_1, n_2 X_2, n_3 X_3, \dots, n_k X_k$

A T C G
in DNA
modeling

Multinomial distribution

- ✱ A discrete random variable X is Multinomial if

$$P(X_1 = n_1, X_2 = n_2, \dots, X_k = n_k) = \frac{N!}{n_1! n_2! \dots n_k!} p_1^{n_1} p_2^{n_2} \dots p_k^{n_k}$$

where $N = n_1 + n_2 + \dots + n_k$

- ✱ The event of throwing k-sided die to see the probability of getting $n_1 X_1, n_2 X_2, n_3 X_3 \dots$

ILLINOIS?

$$\frac{8!}{3!2!1!1!1!}$$

↑ ↑

I L

Multinomial distribution

✱ Examples

- ✱ If we roll a six-sided die N times, how many of each value will we see?
- ✱ What are the counts of N independent and identical distributed trials?
- ✱ This is very widely used in genetics

Multinomial distribution: die example

- ✱ What is the probability of seeing 1 one, 2 twos, 3 threes, 4 fours, 5 fives and 0 sixes in 15 rolls of a fair six-sided die?

Discrete uniform distribution

- ✱ A discrete random variable X is uniform if it takes k different values and

$$P(X = x_i) = \frac{1}{k} \quad \text{For all } x_i \text{ that } X \text{ can take}$$

- ✱ For example:

- ✱ Rolling a fair k -sided die
- ✱ Tossing a fair coin ($k=2$)



Discrete uniform distribution

- ✱ Expectation of a discrete random variable X that takes k different values uniformly

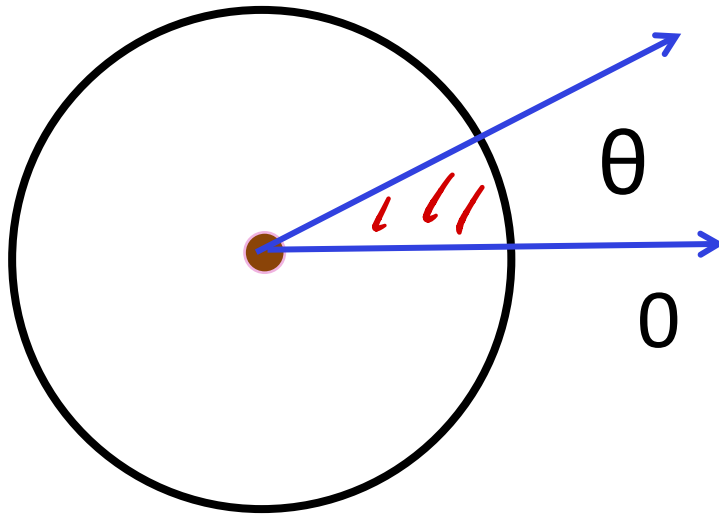
$$E[X] = \frac{1}{k} \sum_{i=1}^k x_i$$

- ✱ Variance of a uniformly distributed random variable X .

$$\text{var}[X] = \frac{1}{k} \sum_{i=1}^k (x_i - E[X])^2$$

Example of a continuous random variable

- ✱ The spinner



$$P(X=k)$$

heads

$$\theta \in (0, 2\pi]$$

- ✱ The sample space for all outcomes is not countable

$$P(\theta = \theta_0) = ?$$

$$P(0 < \theta < \theta_0)$$

~~$$P(\theta = \theta_0)$$~~

Probability density function (pdf)

✱ For a continuous random variable X , the probability that $X=x$ is essentially zero for all (or most) x , so we can't define $P(X = x)$

✱ Instead, we define the probability density function (pdf) over an infinitesimally small

interval dx , $p(x)dx = P(X \in [x, x + dx])$

✱ For $a < b$

$$p(x) = \lim_{dx \rightarrow 0} \frac{P(x_0 < X < x_0 + dx)}{dx}$$
$$\int_a^b p(x)dx = P(X \in [a, b])$$

Properties of the probability density function

- * $p(x)$ **resembles** the probability function of discrete random variables in that
 - * $p(x) \geq 0$ for all x
 - * The probability of X taking all possible values is 1.

$$\int_{-\infty}^{\infty} p(x) dx = 1$$

$= P(X < \infty \text{ \& } X > -\infty) = 1$

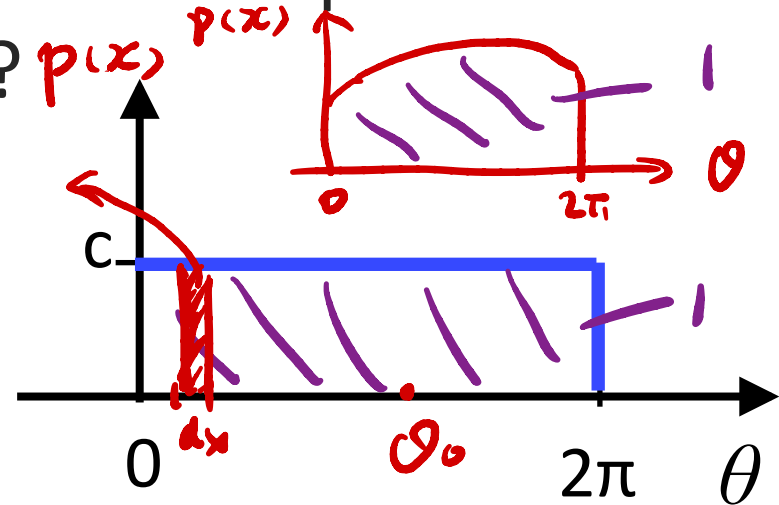
Properties of the probability density function

- ✱ $p(x)$ **differs** from the probability distribution function for a discrete random variable in that
 - ✱ $p(x)$ is not the probability that $X = x$
 - ✱ $p(x)$ can exceed 1

Probability density function: spinner

- Suppose the spinner has equal chance stopping at any position. What's the pdf of the angle θ of the spin position?

$$p(\theta) = \begin{cases} c & \text{if } \theta \in (0, 2\pi] \\ 0 & \text{otherwise} \end{cases}$$

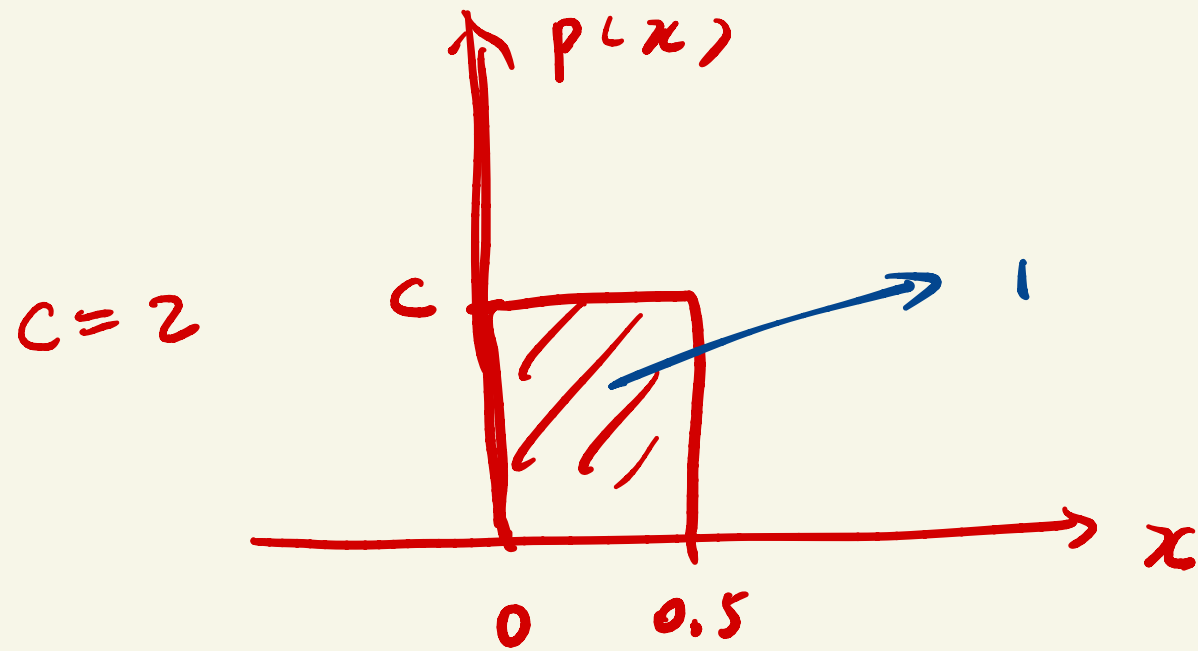


- For this function to be a pdf,

Then

$$\int_{-\infty}^{\infty} p(\theta) d\theta = 1$$

$$P(\theta < \theta_0) = \int_0^{\theta_0} p(x) dx$$



We ask for

$$P(X \in [a, b])$$

$P(X = x_0) \rightarrow$ not for
continuous
RV

Assignments

- ✱ Work on Week5 material
- ✱ Next time: more classic known probability distributions

Additional References

- ✿ Charles M. Grinstead and J. Laurie Snell
"Introduction to Probability"
- ✿ Morris H. Degroot and Mark J. Schervish
"Probability and Statistics"

See you next time

*See
You!*

