

Probability and Random Processes

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Agenda

1. . The notion of expected value
2. Simple computations

The notion of average

The notion of average

The average of a set of numbers x_1, \dots, x_n is

$$\frac{x_1 + \dots + x_n}{n}$$

The notion of average

$$x_1 \rightarrow \textcircled{1} = \underbrace{\frac{1}{n} + \dots + \frac{1}{n}}_{n \text{ terms}}$$

The average of a set of numbers x_1, \dots, x_n is

$$\frac{x_1 + \dots + x_n}{n} = \frac{1}{n}x_1 + \frac{1}{n}x_2 + \dots + \frac{1}{n}x_n$$

This is useful when numbers are equally important, and hence must receive the same weight which is $1/n$.

GPA

- averaging with different weights

2.5	1
5	2
7.5	3

$$\downarrow$$

$$\frac{1+2+3}{3} = 2$$

$$\frac{2.5 \times 1 + 5 \cdot 2 + 3 \cdot 7.5}{2.5 + 5 + 7.5}$$

$$\frac{2.5}{15} \cdot 1 + \frac{5}{15} \cdot 2 + \frac{7.5}{15} \cdot 3$$

$\frac{1}{6}$ $\frac{1}{3}$ $\frac{1}{2}$

The notion of average

The average of a set of numbers x_1, \dots, x_n is

$$\frac{x_1 + \dots + x_n}{n}$$

This is useful when numbers are equally important, and hence must receive the same weight which is $1/n$.

What if the points are not equally important?

x_1, x_2, \dots, x_n
weights w_1, w_2, \dots, w_n

$$w_i \geq 0 \quad 1 \leq i \leq n$$

$$w_1 + w_2 + \dots + w_n = 1$$

$w_1 x_1 + w_2 x_2 + \dots + w_n x_n$ is a weighted average
of x_1, \dots, x_n with weights w_1, \dots, w_n .

Expectation of a random variable

Expected value or Expectation random value.

X discrete random variable, taking values

x_1, x_2, \dots, x_n with p.m.f p_1, \dots, p_n

one-number summary of
the values of X ?

X	x_1	x_2	\dots	x_n
$P(X=x_i)$	p_1	p_2	\dots	p_n

naïve suggestion

$$\frac{x_1 + x_2 + \dots + x_n}{n}$$

Bernoulli

X	0	1
	0.01	0.99

Better suggestion

$$p_1 x_1 + p_2 x_2 + \dots + p_n x_n$$

naïve suggestion : $\frac{0+1}{2} = \frac{1}{2}$

better suggestion : $0.01 \cdot 0 + 0.99 \cdot 1 = 0.99$

Expectation of a random variable

Definition

For a discrete random variable X with values x_1, x_2, \dots, x_n obtained with probabilities p_1, \dots, p_n . Then the *expected value* of X is defined by

$$\mathbb{E}[X] = p_1 x_1 + \dots + p_n x_n.$$

$$\mathbb{E}[X] = \sum_{i=1}^n x_i \cdot \mathbb{P}(X = x_i) = \sum_{i=1}^n x_i \cdot p_i$$

X	x_1	x_2	\dots	x_n
$\mathbb{P}(X=x_i)$	p_1	p_2	\dots	p_n

$\mathbb{E}[X] = p_1 x_1 + p_2 x_2 + \dots + p_n x_n$

Example

Example

A fair die is rolled. Suppose that X is the number shown. Compute $\mathbb{E}[X]$.

Clearly X takes value $1 \leq n \leq 6$, each with probability $1/6$. From here we have:

$$\mathbb{E}[X] = \frac{1}{6} \sum_{n=1}^6 n = \frac{21}{6} = 3.5$$

$$\begin{aligned} & 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + \dots + 6 \cdot \frac{1}{6} \\ &= \frac{1}{6} [1 + 2 + 3 + \dots + 6] \\ &= \frac{21}{6} = 3.5 \end{aligned}$$

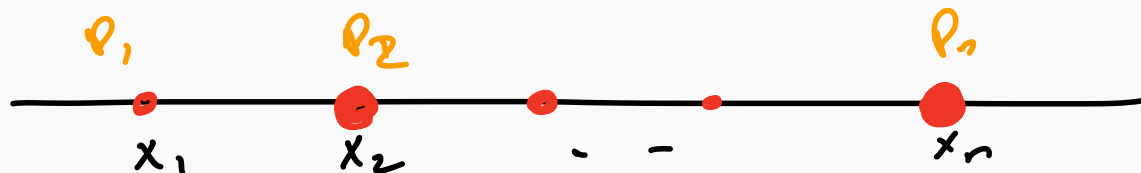
	1	2	3	4	5	6
	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$

Physical interpretation of the expected value

real-valued
random variable



distribution of
mass on the real axis



$x_1 < x_2 < \dots < x_n$
$p_1 \quad p_2 \quad \dots \quad p_n$

Does the notion of expected value
have a physical interpretation?

Center of gravity

Center of mass

probability density
function



$$\text{Center of mass} = \frac{m_1 x_1 + m_2 x_2 + \dots + m_n x_n}{m_1 + m_2 + \dots + m_n}$$

$$M = m_1 + m_2 + \dots + m_n$$

$$= \frac{m_1}{M} x_1 + \frac{m_2}{M} x_2 + \dots + \frac{m_n}{M} x_n$$

Physical interpretation of the expected value

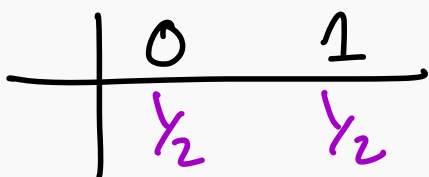
The center of mass of a finite set of points:

Physical interpretation of the expected value

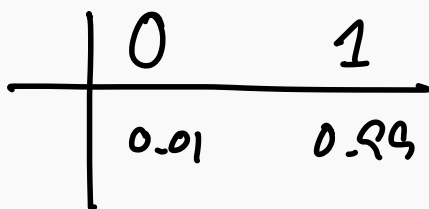
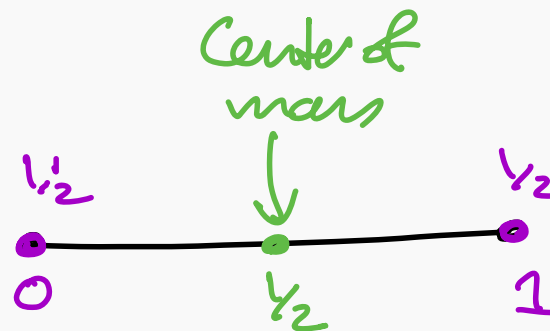
The center of mass of a finite set of points:

Put weights m_1, \dots, m_n at locations x_1, \dots, x_n . Then the center of mass is at

$$\frac{m_1 x_1 + \dots + m_n x_n}{m_1 + \dots + m_n}.$$



$$\frac{1}{2} = \frac{1}{2} \cdot 0 + \frac{1}{2} \cdot 1$$



$$0.01 + 0.99 \times 1 = 0.99$$



Expected value of Bernoulli random variables

X Bernoulli
with parameter p

X	0	1
	$1-p$	p

$$E[X] = \sum_{i=1}^n p_i x_i = (1-p) \times 0 + p \times 1 \\ = p$$

Expected value of Bernoulli random variables

Example

Let X be a Bernoulli random variable with parameter p . Then

$$\mathbb{E}[X] = p \cdot 1 + (1 - p) \cdot 0 = p.$$

Expected value of binomial random variables

X binomial RV with parameters (n, p)

X counts the number of successes in n Bernoulli trials

X takes values $0, 1, \dots, n$. $P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$

Compute $E[X]$.

$$E[X] = \sum_{k=0}^n k \cdot P(X=k)$$

X	0	1	2	...	k	...	n
	$(1-p)^n$				$\binom{n}{k} p^k (1-p)^{n-k}$		p^n

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

$$= \sum_{k=1}^n k \cdot \frac{n!}{k! (n-k)!} \cdot p^k (1-p)^{n-k}$$

$1 \times 2 \times \dots \times k$ (under $k!$)

$(n-1)!$ (under $(n-k)!$)

$$\begin{aligned}
 &= \sum_{k=1}^n \frac{\overbrace{1 \times 2 \times \dots \times (n-1) \times n}^{n!}}{(k-1)! (n-k)!} p^k (1-p)^{n-k} \\
 &= \sum_{k=1}^n \frac{(n-1)! \times n}{(k-1)! (n-k)!} p^k (1-p)^{n-k} \quad \begin{array}{l} n-k+k-1 = n-1 \\ (n-1) - (k-1) = n-k \end{array} \\
 &= n \sum_{k=1}^n \binom{n-1}{k-1} p^k (1-p)^{n-k}
 \end{aligned}$$

$$(a+b)^m = \sum_{j=0}^m \binom{m}{j} a^j b^{m-j}$$

$$\begin{aligned}
 &= n \sum_{j=0}^{n-1} \binom{n-1}{j} p^{j+1} (1-p)^{(n-1)-j} \quad \begin{array}{l} k-1=j \\ k=j+1 \\ n-k=n-j-1 \end{array} \\
 &= np \sum_{j=0}^{n-1} \binom{n-1}{j} p^j (1-p)^{(n-1)-j} \\
 &= np \cdot \underbrace{(p + 1-p)}_1^{n-1} = np.
 \end{aligned}$$

X binomial random variable with parameters (n, p)

Then	$E[X] = np.$	
Ex. Coin	H	$\frac{1}{4}$ success 100 times
	T	$\frac{3}{4}$
	$p = \frac{1}{4}$	$n = 100$

Expected value of binomial random variables

Example

Let X be a Binomial random variable with the parameters (n, p) . Find $\mathbb{E}[X]$ in terms of n and p .

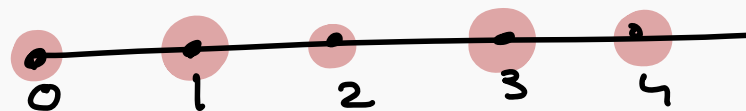
Formula for specific random variables

Theorem

Suppose X is a random variable which only takes values $0, 1, 2, \dots$. Then the expected value of X is given by

$$\underline{\mathbb{E}[X] = \sum_{i=1}^{\infty} \mathbb{P}[X \geq i].}$$

$$\mathbb{E}[X] = \sum_{j=0}^{\infty} j \cdot p_j$$



$$= \cancel{0 \cdot p_0} + \cancel{1} \cdot p_1 + \textcolor{blue}{2} \cdot p_2 + \textcolor{yellow}{3} \cdot p_3 + \textcolor{green}{4} \cdot p_4 + \dots$$

0	1	2	3	...
p_0	p_1	p_2	p_3	...

$$= \begin{array}{l} \textcolor{red}{p_1} + \textcolor{blue}{p_2} + \textcolor{yellow}{p_3} + \textcolor{green}{p_4} + \dots = \mathbb{P}(X \geq 1) \\ + \textcolor{blue}{p_2} + \textcolor{yellow}{p_3} + \textcolor{green}{p_4} + \dots = \mathbb{P}(X \geq 2) \\ + \textcolor{yellow}{p_3} + \textcolor{green}{p_4} + \dots = \mathbb{P}(X \geq 3) \\ + \textcolor{green}{p_4} + \dots = \mathbb{P}(X \geq 4) \\ \vdots \end{array}$$

$p_0 + p_1 + p_2 + \dots = 1$

Example

Let X have the geometric distribution with parameter p , i.e.

$$\mathbb{P}[X = j] = p(1 - p)^{j-1},$$

for $j \geq 1$. Compute $\mathbb{E}[X]$.

$$\mathbb{E}(X) = 1 \cdot p + 2(1-p)p + 3(1-p)^2p + 4(1-p)^3p + \dots$$

j	1	2	3	4	...
$\mathbb{P}[X=j]$	p	$(1-p)p$	$(1-p)^2p$	$(1-p)^3p$...

Can we do better?

Values of $X = \{1, 2, 3, \dots\} \subseteq \{0, 1, 2, 3, \dots\}$

$$\begin{aligned} \mathbb{E}[X] &= \sum_{j=1}^{\infty} \underbrace{\mathbb{P}[X \geq j]}_{(1-p)^{j-1}} = \sum_{j=1}^{\infty} (1-p)^{j-1} = \sum_{\substack{k=0 \\ j-1=k}}^{\infty} (1-p)^k \stackrel{1-p=q}{=} 1 + q + q^2 + \dots \\ &= \frac{1}{1-q} = \frac{1}{p} \end{aligned}$$

Summary

X random variable with a geometric distribution with parameter p , then

$$E[X] = 1/p.$$

X : first success in a Bernoulli trials.
prob of success = p .

p close to 1. $p = 0.8$ $E[X] = \frac{1}{p} = \frac{1}{8/10}$

p close to 0 $p = 0.1$

$$E[X] = \frac{1}{1/10} = 10.$$

$$\begin{array}{c|ccc} & 1 & 2 & 3 \dots \\ \hline & 0.8 & \frac{16}{100} & \end{array}$$

$$\begin{array}{c|ccc} & 1 & 2 & 3 \\ \hline & \frac{1}{10} & \frac{9}{10} \cdot \frac{1}{10} & \\ & & \frac{9}{100} & \approx \frac{1}{10} \end{array}$$

$$X = Y_1 + Y_2$$

$$E[X] = E[Y_1 + Y_2]$$