

# Variance, Standard Deviation of Discrete and Continuous Random Variables

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Lesson 5

# Contents

- ❑ Discrete Random Variables
  - ❑ Expected Value
  - ❑ Variance
  - ❑ Standard Deviation
- ❑ Continuous Random Variable
  - ❑ Expected Value
  - ❑ Variance
  - ❑ Standard Deviation

**DISCRETE**

# Expected Value

**Definition:** Suppose  $X$  is a discrete random variable that takes values  $x_1, x_2, \dots, x_n$  with probabilities  $p(x_1), p(x_2), \dots, p(x_n)$ .

The expected value of  $X$  is denoted  $E(X)$  and defined by:

$$E(X) = p(x_1) \cdot x_1 + p(x_2) \cdot x_2 + \dots + p(x_n) \cdot x_n$$

# Properties of Expected Value

1. If  $X$  and  $Y$  are random variables on a sample space  $\Omega$  then:

$$E(X + Y) = E(X) + E(Y)$$

2. If  $a$  and  $b$  are constants then:

$$E(aX + b) = aE(X) + b$$

**Example 1:** Roll two dice and let  $X$  be the sum. Find  $E(X)$ .

# Properties of Expected Value

**Example 2:** (For infinite random variables the “mean” (expected value) does not always exist.) Suppose  $X$  has an infinite number of values according to the following table.

	$x_1$	$x_2$	$x_3$	...	$x_k$	
values $x$ :	2	$2^2$	$2^3$	...	$2^k$	...
pmf $p(x)$ :	$1/2$	$1/2^2$	$1/2^3$	...	$1/2^k$	...

# Expected values of functions

If  $X$  is a discrete random variable taking values  $x_1, x_2, \dots$  and  $h$  is a function the  $h(X)$  is a new random variable. Its expected value is

$$E(h(X)) = \sum ( h(x_j) \cdot p(x_j) )$$

**Example 3:** Let  $X$  be the value of a roll of one die and let  $Y = X^2$ . Find  $E(Y)$ .

**Example 4:** Roll two dice and let  $X$  be the sum. Suppose the payoff function is given by  $Y = X^2 - 6X + 1$ .

# Variance

**Definition:** If  $X$  is a discrete random variable with mean  $E(X) = \mu$ , then the variance of  $X$  is defined by:

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

And is denoted by  $\sigma^2$ .

“ Is a measure of how much the probability mass is spread out around this center.”



# Standard Deviation

**Definition:** If  $X$  is a discrete random variable, then the standard deviation of  $X$  is defined by:

$$\sigma = (\text{Var}(X))^{\frac{1}{2}}$$

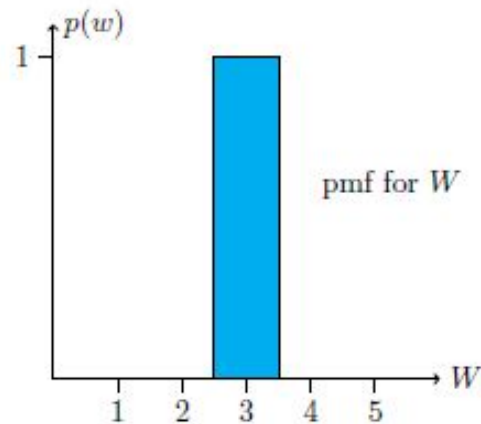
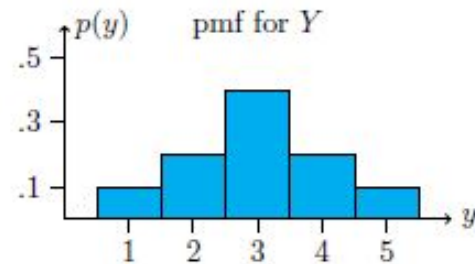
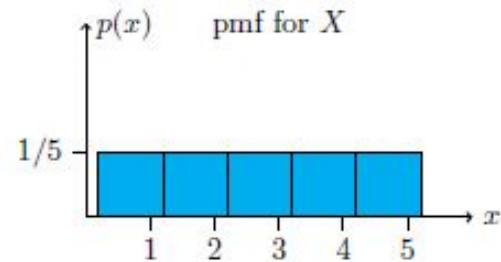
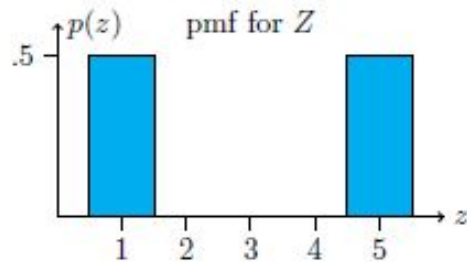
**Example 5:** Compute the mean (expected), variance and standard deviation of the random variable  $X$  with the following table of values and probabilities.

value $x$	1	3	5
pmf $p(x)$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$

**Example 6:** For each random variable  $X$ ,  $Y$ ,  $Z$ , and  $W$  plot the pmf and compute the mean (expected value) and variance.

(i)	<table><tr><td>value <math>x</math></td><td>1</td><td>2</td><td>3</td><td>4</td><td>5</td></tr><tr><td>pmf <math>p(x)</math></td><td>1/5</td><td>1/5</td><td>1/5</td><td>1/5</td><td>1/5</td></tr></table>	value $x$	1	2	3	4	5	pmf $p(x)$	1/5	1/5	1/5	1/5	1/5
value $x$	1	2	3	4	5								
pmf $p(x)$	1/5	1/5	1/5	1/5	1/5								
(ii)	<table><tr><td>value <math>y</math></td><td>1</td><td>2</td><td>3</td><td>4</td><td>5</td></tr><tr><td>pmf <math>p(y)</math></td><td>1/10</td><td>2/10</td><td>4/10</td><td>2/10</td><td>1/10</td></tr></table>	value $y$	1	2	3	4	5	pmf $p(y)$	1/10	2/10	4/10	2/10	1/10
value $y$	1	2	3	4	5								
pmf $p(y)$	1/10	2/10	4/10	2/10	1/10								
(iii)	<table><tr><td>value <math>z</math></td><td>1</td><td>2</td><td>3</td><td>4</td><td>5</td></tr><tr><td>pmf <math>p(z)</math></td><td>5/10</td><td>0</td><td>0</td><td>0</td><td>5/10</td></tr></table>	value $z$	1	2	3	4	5	pmf $p(z)$	5/10	0	0	0	5/10
value $z$	1	2	3	4	5								
pmf $p(z)$	5/10	0	0	0	5/10								
(iv)	<table><tr><td>value <math>w</math></td><td>1</td><td>2</td><td>3</td><td>4</td><td>5</td></tr><tr><td>pmf <math>p(w)</math></td><td>0</td><td>0</td><td>1</td><td>0</td><td>0</td></tr></table>	value $w$	1	2	3	4	5	pmf $p(w)$	0	0	1	0	0
value $w$	1	2	3	4	5								
pmf $p(w)$	0	0	1	0	0								

- $\text{Var}(X) = 2$
- $\text{Var}(Y) = 1.2$
- $\text{Var}(Z) = 4$
- $\text{Var}(W) = 0$



# Properties of Variance

- If  $X$  and  $Y$  are independent then  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$
- For constants  $a$  and  $b$ , then:  $\text{Var}(a \cdot X + b) = a^2 \text{Var}(X)$
- $\text{Var}(X) = E(X^2) - E(X)^2$

**Example 7:** Suppose  $X$  and  $Y$  are independent and  $\text{Var}(X) = 3$  and  $\text{Var}(Y) = 5$ .

Find:

(i)  $\text{Var}(X + Y)$

(ii)  $\text{Var}(3X + 4)$

(iii)  $\text{Var}(X + X)^*$

(iv)  $\text{Var}(X + 3Y)$

# Summary

Expected Value:		Variance:
Synonyms:	mean, average	
Notation:	$E(X), \mu$	$\text{Var}(X), \sigma^2$
Definition:	$E(X) = \sum_j p(x_j)x_j$	$E((X - \mu)^2) = \sum_j p(x_j)(x_j - \mu)^2$
Scale and shift:	$E(aX + b) = aE(X) + b$	$\text{Var}(aX + b) = a^2\text{Var}(X)$
Linearity:	(for any $X, Y$ ) $E(X + Y) = E(X) + E(Y)$	(for $X, Y$ independent) $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$
Functions of $X$ :	$E(h(X)) = \sum p(x_j) h(x_j)$	
Alternative formula:		$\text{Var}(X) = E(X^2) - E(X)^2 = E(X^2) - \mu^2$

**CONTINUOUS**

# Expected Value

**Definition:** Let  $X$  be a continuous random variable with range  $[a, b]$  and probability density function  $f(x)$ . The expected value of  $X$  is defined by:

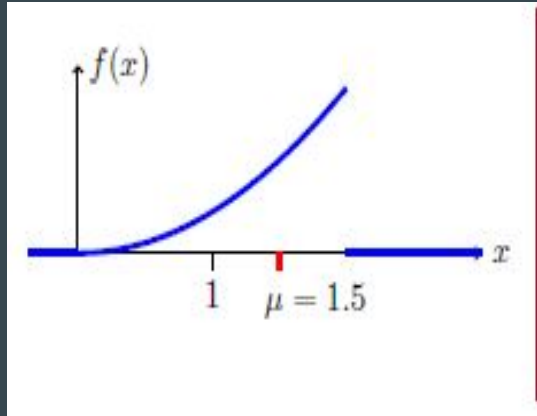
$$E(X) = \int_a^b x f(x) dx$$

**Example 8:** Let  $X$  have range  $[0, 2]$  and density  $f(x) = 3x^2/8$ . Find  $E(X)$ .

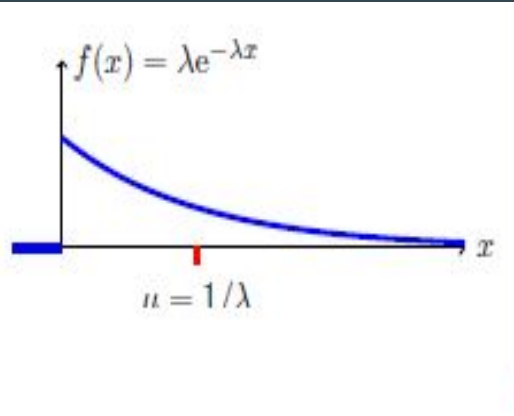
**Example 9:** The range of  $X$  is  $[0, \infty)$  and its pdf is  $f(x) = \lambda e^{-\lambda x}$

**Example 10:** The range of  $Z$  is  $(-\infty, \infty)$  and its pdf is  $f(x) = e^{-Z^2/2}/\sqrt{2\pi}$

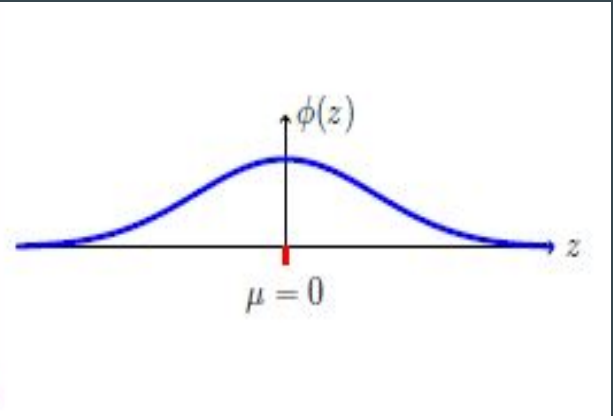
# Expected Value



Example 8



Example 9



Example 10



# Properties of Expected Value

The properties of  $E(X)$  for continuous random variables are the same as for discrete ones:

1. If  $X$  and  $Y$  are random variables on a sample space  $\Omega$  then:

$$E(X + Y) = E(X) + E(Y)$$

2. If  $a$  and  $b$  are constants then:

$$E(a \cdot X + b) = a \cdot E(X) + b$$

# Expectation of Functions of X

This works exactly the same as the discrete case. If  $h(x)$  is a function then  $Y = h(X)$  is a random variable and

$$E(Y) = E(h(X)) = \int h(x) \cdot f(x) \, dx$$

**Example 11:** Find  $E(X^2)$  if the range of  $X$  is  $[0, \infty)$  and its pdf is  $f(x) = \lambda e^{-\lambda x}$

# Variance

**Definition:** Let  $X$  be a continuous random variable with mean. The variance of  $X$  is:

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

**Properties:**

- If  $X$  and  $Y$  are independent then:  $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y)$
- For constants  $a$  and  $b$ :  $\text{Var}(aX + b) = a^2 \text{Var}(X)$
- $\text{Var}(X) = E(X^2) - E(X)^2$

**Example 12:** Find  $\text{Var}(X)$ .  $X$  has range  $[0, 1]$  and density  $f(x) = 1$ .

# Dessert

We roll two standard 6-sided dice. You win \$1000 if the sum is 2 and lose \$100 otherwise. How much do you expect to win on average per trial?

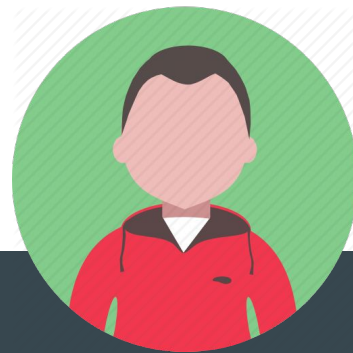
# Math Team



Christian Córdova

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Bachelor in Mathematics



José Castro

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Bachelor in Computer  
Science

Statistics

Probability

Deviation

Mode

Median

distribution

Standard

Mean

population

Z-score

standard

event

stem-leaf

union

Sample

Normal

independence

intersection  
box-plot  
observation

randomness

dotplot

Skewed

Venn-Diagram

histogram

conditional

Random

chance

bargraph

probability-tree

pie-chart