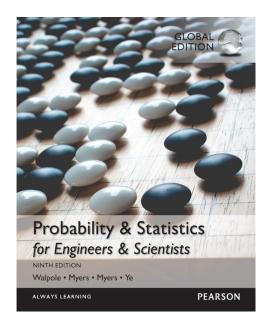
Chapter 4 Mathematical Expectation – part 1

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Outline

- Mean of a Random Variable
- Variance and Covariance of Random Variables
- Means and Variances of Linear Combinations of Random Variables
- Chebyshev's Theorem



4.1 Mean of a Random Variable



Question!

- Consider a casino game in which the probability of losing \$1 per game is 0.8 and the probability 0.2 win \$2 per game.
- The gain or loss of a gambler who plays this game only a few times depends on his "luck" more than anything else.
- But, if a gambler decides to play the game a large number of times, his loss or gain depends more on "the number of plays" than on his luck.
- "Expected" total gain when playing this game n times = ??

$$(0.8)n \cdot (-1) + (0.2)n \cdot 2 = (-0.4)n.$$



Average ?

- In Ch. 1, we discussed sample mean; arithmetic mean of the data
- If two coins are tossed 16 times and X is the number of heads that occur per toss, then the values of X are 0, 1, and 2.
- Suppose that the experiment yields no heads, one head, and two heads a total of 4, 7, and 5 times, respectively.
 The average number of heads per toss of the two coins is then

$$\frac{(0)(4) + (1)(7) + (2)(5)}{16} = 1.06.$$



 Let us now restructure our computation for the average number of heads so as to have the following equivalent form:

$$(0) \left(\frac{4}{16}\right) + (1) \left(\frac{7}{16}\right) + (2) \left(\frac{5}{16}\right) = 1.06.$$
mean or expectation of $X(\mu_x)$

the fractions of the total tosses resulting in 0, 1, and 2 heads, respectively



Mean and Expectation*

- This average value is called **mean of the** random variable X, or the **mean of the probability distribution of X**, and write it as μ_x or simply as μ when it is clear to which random variable we refer
- It is also common to refer to this mean as the mathematical expectation, or expected value of random variable X, and denote it as E(X).



Assuming that 1 fair coin was tossed twice, we find that the sample space for our experiment is

$$S = \{HH, HT, TH, TT\}.$$

Since the 4 sample points are all equally likely, it follows that

$$P(X = 0) = P(TT) = \frac{1}{4}, \quad P(X = 1) = P(TH) + P(HT) = \frac{1}{2},$$

and

$$P(X = 2) = P(HH) = \frac{1}{4},$$

where a typical element, say TH, indicates that the first toss resulted in a tail followed by a head on the second toss. Now, these probabilities are just the relative frequencies for the given events in the long run. Therefore,

$$\mu = E(X) = (0)\left(\frac{1}{4}\right) + (1)\left(\frac{1}{2}\right) + (2)\left(\frac{1}{4}\right) = 1.$$

This result means that a person who tosses 2 coins over and over again will, on the average, get 1 head per toss. $\mu = E(X) = \sum_x x f(x)$



Expectation

Definition 4.1:

Let X be a random variable with probability distribution f(x). The **mean**, or **expected value**, of X is

$$\mu = E(X) = \sum_{x} x f(x)$$

if X is discrete, and

$$\mu = E(X) = \int_{-\infty}^{\infty} x f(x) \ dx$$

if X is continuous.

In the case of continuous random variables, the definition of an expected value is essentially the same with summations replaced by integrations.



Example 4.1

- A box containing 7 components is sampled by a quality inspector; the box contains 4 good component and 3 defective components.
- A sample of 3 is taken by the inspector. Find the expected value of the number of good components in this sample.

Solution:

 Let X be represent the number of good components in the sample. The probability distribution of X is

$$f(x) = \frac{\binom{4}{x}\binom{3}{3-x}}{\binom{7}{3}}, \qquad x = 0, 1, 2, 3.$$

$$\mu = E(X) = \sum_{x} x f(x) = 0 \cdot \frac{1}{35} + 1 \cdot \frac{12}{35} + 2 \cdot \frac{18}{35} + 3 \cdot \frac{4}{35} = \frac{12}{7}.$$



Meaning:

Thus, if a sample of size 3 is selected at random over and over again from a box of 4 good components and 3 defective components, it would contain, on average, 12/7 (≈ 1.7) good components.



Question!

 In a gambling game a man is paid \$5 if he gets all heads or all tails when three coin are tossed, and he will pay out \$3 if either one or two heads show. What is his expected gain?





Solution:

- The sample space for the possible outcomes when three coins are tossed simultaneously is
- S = {HHH, HHT, HTH, THH, HTT, THT, TTH, TTT}.
- The random variable of interest is Y, the amount the gambler can win; and the possible values of Y are
 - \$5 if event E₁ = {HHH, TTT} occurs and
 - \$-3 if event E2 = {HHT, HTH, THH, HTT, THT, TTH} occurs,
- that is, the probability function of Y is given by

$$f(y) = P(X = y) = \begin{cases} \frac{1}{4}, & y = 5; \\ \frac{3}{4}, & y = -3; \\ 0, & \text{elsewhere.} \end{cases} \quad \mu = E(Y) = 5 \cdot \frac{1}{4} + (-3) \cdot \frac{3}{4} = -1.$$



- Example 4.3
 - Let X be the random variable that denotes the life in hours of a certain electronic device. The probability density function is

$$f(x) = \begin{cases} \frac{20,000}{x^3}, & x > 100, \\ 0, & \text{elsewhere.} \end{cases}$$

- Find the expected life of this type of device.

Solution:
$$\mu = E(X) = \int_{-\infty}^{\infty} x f(x) \ dx$$

$$\mu = E(X) = \int_{100}^{\infty} x \frac{20,000}{x^3} dx = \int_{100}^{\infty} \frac{20,000}{x^2} dx = 200.$$

Therefore, we can expect this type of device to last, on average, 200 hours.



Expectation of Function

• For the following probability mass function for X, calculate $E[X^2]$

$$p(0) = .2$$
, $p(1) = .5$, $p(2) = .3$

Let a random variable $Y = X^2$,

$$p_Y(0) = P\{Y = 0^2\} = .2$$

 $p_Y(1) = P\{Y = 1^2\} = .5$
 $p_Y(4) = P\{Y = 2^2\} = .3$

$$\mu = E(X) = \sum_{x} x f(x)$$

$$E[X^2] = E[Y] = 0(.2) + 1(.5) + 4(.3) = 1.7$$



Theorem 4.1

Let X be a random variable with probability distribution f(x). The expected value of the random variable g(X) is

$$\mu_{g(X)} = E[g(X)] = \sum_{x} g(x)f(x)$$

if X is discrete, and

$$\mu_{g(X)} = E[g(X)] = \int_{-\infty}^{\infty} g(x)f(x) dx$$

if X is continuous.



Expectation of Function: Example

For the following probability mass function for X, calculate $E[X^2]$

$$p(0) = .2, \quad p(1) = .5, \quad p(2) = .3$$

Using the Theorem 4.1,
$$\mu_{g(X)} = E[g(X)] = \sum_x g(x)f(x)$$

$$E[X^2] = 0^2(0.2) + (1^2)(0.5) + (2^2)(0.3) = 1.7$$



Example 4.4

 Suppose that the number of cars X that pass through a car wash between 4:00 P.M. and 5:00 P.M. on any sunny Friday has the following probability dist.:

x	4	5	6	7	8	9
P(X=x)	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{6}$	$\frac{1}{6}$

- Let g(X) = 2X-1 represent the amount of money, in dollars, paid to the attendant by the manager.
- Find the attendant's expected earnings for this particular time period.



x	4	5	6	7	8	9
P(X=x)	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{6}$	$\frac{1}{6}$

Solution:

By theorem, the attendant can expect to receive

$$E[g(X)] = E(2X - 1) = \sum_{x=4}^{9} (2x - 1)f(x)$$

$$\mu_{g(X)} = E[g(X)] = \sum_{x=4}^{9} g(x)f(x)$$

$$= (7)\left(\frac{1}{12}\right) + (9)\left(\frac{1}{12}\right) + (11)\left(\frac{1}{4}\right) + (13)\left(\frac{1}{4}\right) + (15)\left(\frac{1}{6}\right) + (17)\left(\frac{1}{6}\right) = \$12.67.$$



- Example 4.5:
 - Let X be a random variable with density function

$$f(x) = \begin{cases} \frac{x^2}{3}, & -1 < x < 2, \\ 0, & \text{elsewhere.} \end{cases}$$

- Find the expected value of g(X) = 4X + 3.
- Solution:

By Theorem 4.1, we have

$$\mu_{g(X)} = E[g(X)] = \int_{-\infty}^{\infty} g(x)f(x) \ dx$$

$$E(4X+3) = \int_{-1}^{2} \frac{(4x+3)x^2}{3} dx = \frac{1}{3} \int_{-1}^{2} (4x^3 + 3x^2) dx = 8.$$



Definition 4.2:

For joint probability distribution f(x, y),

Let X and Y be random variables with joint probability distribution f(x, y). The mean, or expected value, of the random variable g(X, Y) is

$$\mu_{g(X,Y)} = E[g(X,Y)] = \sum_{x} \sum_{y} g(x,y) f(x,y)$$

if X and Y are discrete, and

$$\mu_{g(X,Y)} = E[g(X,Y)] = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x,y)f(x,y) dx dy$$

if X and Y are continuous.



Example 4.7: Find E(Y/X) for the density function

$$f(x,y) = \begin{cases} \frac{x(1+3y^2)}{4}, & 0 < x < 2, \ 0 < y < 1, \\ 0, & \text{elsewhere.} \end{cases}$$

Solution: We have

$$E\left(\frac{Y}{X}\right) = \int_0^1 \int_0^2 \frac{y(1+3y^2)}{4} \ dxdy = \int_0^1 \frac{y+3y^3}{2} \ dy = \frac{5}{8}.$$



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4.2 Variance andCovariance of Random Variables (1)



Quantifying Variability: Variance

- The **variance** of a random variable is a measure of its statistical dispersion (분산, 이산), indicating how its possible values are spread around the expected value.
 - the expected value shows the location of the distribution,
 - the variance indicates the variability of the values.

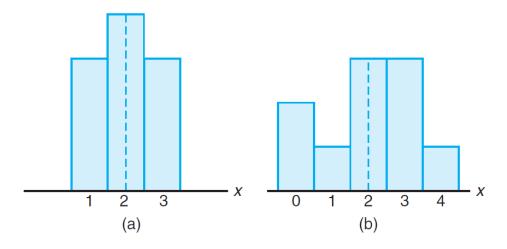


Figure 4.1: Distributions with equal means and unequal dispersions.



Variance

Definition 4.3: variance of random variable X

Let X be a random variable with probability distribution f(x) and mean μ . The variance of X is

$$\sigma^2 = E[(X - \mu)^2] = \sum_x (x - \mu)^2 f(x), \quad \text{if } X \text{ is discrete, and}$$

$$\sigma^2 = E[(X - \mu)^2] = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) \, dx, \quad \text{if } X \text{ is continuous.}$$

The positive square root of the variance, σ , is called the **standard deviation** of X.

- Variation : σ^2 Var(X)
- The positive square root of the variance, σ, is called the standard deviation of X.



$$\sigma = \sqrt{\mathsf{Var}(X)}.$$

Example 4.8

 Let the random variable X represent the number of automobiles that are used for official business purposes on any given workday. The probability distribution for company A is

and that for company B is

 Show that the variance of the probability distribution for company B is greater than that for company A.



company B

x	1	2	3
f(x)	0.3	0.4	0.3

Solution:

$$\mu = E(X) = \sum_{x} x f(x)$$

$$\mu_A = 1 \cdot 0.3 + 2 \cdot 0.4 + 3 \cdot 0.3 = 2;$$

$$\mu_B = 0 \cdot 0.2 + 1 \cdot 0.1 + 2 \cdot 0.3 + 3 \cdot 0.3 + 4 \cdot 0.1 = 2;$$

$$\sigma_A^2 = \sum_{x=1}^2 (x-2)^2 f_A(x) = 0.6;$$

$$\sigma^{2} = E[(X - \mu)^{2}] = \sum_{x} (x - \mu)^{2} f(x)$$

$$\sigma_B^2 = \sum_{x=0}^4 (x-2)^2 f_B(x) = 1.6.$$



Remember!

- Theorem 4.2
 - The variance of a random variable X is

$$\sigma^2 = E(X^2) - \mu^2.$$

Prove it!



proof

- Proof:
 - For the discrete case we can write:

• Definition 4.3
$$\sigma^{2} = \sum_{x} (x - \mu)^{2} f(x) = \sum_{x} (x^{2} - 2\mu x + \mu^{2}) f(x)$$

$$= \sum_{x} x^{2} f(x) - 2\mu \sum_{x} x f(x) + \mu^{2} \sum_{x} f(x).$$

$$= \sum_{x} x^{2} f(x) - 2\mu \cdot \mu + \mu^{2} \cdot 1.$$
• Definition 4.1
$$= \sum_{x} x^{2} f(x) - \mu^{2}$$

$$= \sum_{x} x^{2} f(x) - \mu^{2}.$$

Theorem 4.1

$$\mu_{g(X)} = E[g(X)] = \sum_x g(x) f(x)$$

* For the continuous case the proof is step by step the same, with summations(Σ) replaced by integrations(J).



Example 4.9

 Let the random variable X represent the number of defective parts for a machine when 3 parts are sampled from a production line and tested. The following is the probability distribution of X.

• Solution: $\sigma^2 = E(X^2) - \mu^2$.

$$\mu = 0 \cdot 0.51 + 1 \cdot 0.38 + 2 \cdot 0.10 + 3 \cdot 0.01 = 0.61;$$

$$E(X^2) = 0 \cdot 0.51 + 1 \cdot 0.38 + 4 \cdot 0.10 + 9 \cdot 0.01 = 0.87;$$

$$\sigma^2 = E(X^2) - \mu^2 = 0.87 - 0.61^2 = 0.4979.$$



Example 4.10

 The weekly demand for a drinking-water product, in thousands of liters, from a local chain of efficiency stores is a continuous random variable X having the probability density

$$f(x) = \begin{cases} 2(x-1), & 1 < x < 2, \\ 0, & \text{elsewhere.} \end{cases}$$

Find the mean and variance of X

$$\mu = E(X) = 2 \int_{1}^{2} x(x-1) dx = \frac{5}{3}$$

and

$$\sigma^2 = E(X^2) - \mu^2.$$

$$E(X^2) = 2\int_1^2 x^2(x-1) \ dx = \frac{17}{6}.$$

Therefore,

$$\sigma^2 = \frac{17}{6} - \left(\frac{5}{3}\right)^2 = \frac{1}{18}.$$

Next important Theorem!

- Let X be a random variable with probability function f(x).
- The variance of the random variable g(X) is
- Theorem 4.3

Let X be a random variable with probability distribution f(x). The variance of the random variable g(X) is

$$\sigma_{g(X)}^2 = E\{[g(X) - \mu_{g(X)}]^2\} = \sum_{x} [g(x) - \mu_{g(X)}]^2 f(x)$$

if X is discrete, and

$$\sigma_{g(X)}^2 = E\{[g(X) - \mu_{g(X)}]^2\} = \int_{-\infty}^{\infty} [g(x) - \mu_{g(X)}]^2 f(x) \, dx$$

if X is continuous.



Example 4.11

• Calculate the variance of g(X) = 2X + 3, where X is a random variable with probability distribution

Theorem 4.1

Solution:

Theorem 4.3
$$\mu_{g(X)} = E[g(X)] = \sum_{x} g(x)f(x)$$

$$\mu_{2X+3} = E(2X+3) = \sum_{x=0}^{3} (2x+3)f(x) = 6.$$

$$\sigma_{2X+3}^2 = E\{[(2X+3) - \mu_{2X+3}]^2\} = E[(2X+3-6)^2]$$

$$= E(4X^2 - 12X + 9) = \sum_{x=0}^{3} (4x^2 - 12x + 9)f(x) = 4.$$

We will study a method to solve this in an easier way in Sec. 4.3.

Example 4.12

Let X be a random variable having the density function given in Example 4.5 on page 135. Find the variance of the random variable g(X) = 4X + 3.

Solution:

$$\sigma_{g(X)}^2 = E\{[g(X) - \mu_{g(X)}]^2\} = \int_{-\infty}^{\infty} [g(x) - \mu_{g(X)}]^2 f(x) \ dx$$

In Example 4.5, we found that $\mu_{4X+3} = 8$. Now, using Theorem 4.3,

$$\sigma_{4X+3}^2 = E\{[(4X+3)-8]^2\} = E[(4X-5)^2]$$

$$= \int_{-1}^2 (4x-5)^2 \frac{x^2}{3} dx = \frac{1}{3} \int_{-1}^2 (16x^4 - 40x^3 + 25x^2) dx = \frac{51}{5}.$$





https://www.psycom.net/bipolar-questions-answers

