MA212 Probability and Statistics

Chapter Four: Expectation and Variance

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1.1 Basic concepts

Expectation of R.V.

Discrete: Assuming the frequency function of r.v. X is p(x), if

$$\sum_{k=1}^{\infty} |x_k| p_k < +\infty,$$

Then

$$E(X) \triangleq \sum_{k=1}^{\infty} x_k p_k = \sum_{k=1}^{\infty} x_k P\{X = x_k\}$$

Continuous: Assuming the probability density function of r.v. X is f(x), if

$$\int_{-\infty}^{\infty} |x| f(x) dx < +\infty$$

Then

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

1.1 Basic concepts (Cont'd)

Expectation of R.V.'s function Y = g(X)

• If X is discrete with frequency function p(x),

when
$$\sum_{k=1}^{\infty} |g(x_k)| \cdot p_k < +\infty$$
, then $E(Y) = E[g(X)] = \sum_{k=1}^{\infty} g(x_k) p_k$

• If X is continuous with density function f(x),

when
$$\int_{-\infty}^{\infty} |g(x)| f(x) dx < \infty$$
, then

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

1.1 Basic concepts (Cont'd)

Expectation of R.V.s' function Z = g(X, Y)

• For the joint frequency function of X and Y is p(i,j). If

$$\sum_{i=1}^{\infty} \sum_{j=1}^{\infty} |g(x_i, y_j)| p_{ij} < +\infty, \text{ then}$$

$$E(Z) = E[g(X,Y)] = \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} g(x_i, y_j) p_{ij}$$

• For the joint density of X and Y is f(x, y). If

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |g(x,y)| f(x,y) dx dy < \infty, \text{ then}$$

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

Note: the formula can be generalized to high-dimensional random variables.

1.2 The properties of expectation

Properties

If
$$a \le X \le b$$
 (a.e), then $a \le E(X) \le b$
If c is constant, then $E(cX) = cE(X)$
If X, Y are r.v., then $E(X+Y) = E(X) + E(Y)$
If X, Y are independent, then $E(XY) = E(X)E(Y)$

Inferences

If
$$X = c$$
 $(a.e)$, then $E(X) = c$
If a_1, a_2, \dots, a_n are constants X_1, X_2, \dots, X_n
are r.v., then
$$E(\sum_{i=1}^n a_i X_i) = \sum_{i=1}^n a_i E(X_i)$$
If X_1, X_2, \dots, X_n are independent, then
$$E(X_1 X_2 \dots X_n) = E(X_1) E(X_2) \dots E(X_n)$$

1.3 Relationship with probability: Inequality

Markov's Inequality

Theorem: If X is a R.V. with $P(X \ge 0) = 1$ and for which E(X) exists, then

$$P\{X \ge t\} \le \frac{E(X)}{t}.$$

Proof: We will prove this for the discrete case. the continuous case is entirely analogous.

$$E(X) = \sum_{x} xp(x) = \sum_{x < t} xp(x) + \sum_{x \ge t} xp(x)$$

All the terms in the sums are nonnegative because X takes on only nonnegative values. Thus

$$E(X) \ge \sum_{x \ge t} xp(x) \ge \sum_{x \ge t} tp(x) = tP\{X \ge t\}.$$

$$\longrightarrow P\{X \ge t\} \le \frac{E(X)}{t}.$$

1.4 Basic concepts (Cont'd)

Variance of R.V.

The calculation of variance is the expected value of $g(X) = (X - E(X))^2$.

$$\operatorname{Var}(X) \stackrel{\Delta}{=} D(X) \stackrel{\Delta}{=} E(X - E(X))^2$$

• Assuming the frequency function of X is p(x), then

$$D(X) = \sum_{k=1}^{\infty} (x_k - E(X))^2 \cdot p_k$$

• Assuming the probability density of X is, then

$$D(X) = \int_{-\infty}^{\infty} (x - E(X))^2 f(x) dx$$

More general, you can use the following equation to calculate variance

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

1.5 The properties of variance

Properties

If
$$X = c(constant)$$
, then $D(X) = 0$

If c is constant, then
$$D(cX) = c^2 D(X)$$

For independent
$$r.v X, Y D(X+Y) = D(X) + D(Y)$$

$$D(X - Y) = D(X) + D(Y)$$

If X and Y are not independent

$$D(X \pm Y) = D(X) + D(Y) \pm 2E[(X - E(X))(Y - E(Y))]$$

1.6 Summary of expectation and variances

Some common distributions

Distribution of X	E(X)	D(X)
$X \sim p(\lambda)$	λ	λ
$X \sim b(n, p)$	np	np(1-p)
$X \sim U(a,b)$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$
$X \sim EXP(1/\theta)$	heta	$ heta^2$
$X \sim N(\mu, \sigma^2)$	μ	σ^2

1.7 Basic concepts (Cont'd)

Covariance of R.V

Definition: If the variance of both X and Y exists, then

$$Cov(X,Y) \stackrel{\triangle}{=} E[(X - E(X))(Y - E(Y))]$$

Cov(X,Y) is the Covariance of X and Y.

$$X, Y$$
 are independent \longrightarrow Cov $(X, Y) = 0$

$$Cov(X,Y) \neq 0 \implies X,Y$$
 are not independent

 \rightarrow X,Y must have a relation

1.8 The properties of Covariance

Properties

- If X, Y are independent, then Cov(X, Y) = 0
- Cov(X,Y) = Cov(Y,X)
- D(X+Y) = D(X) + D(Y) + 2E[(X-E(X))(Y-E(Y))]= D(X) + D(Y) + 2Cov(X,Y)
- For any constant a, bCov(aX, bY) = E[(aX - E(aX))(bY - E(bY))]
- $Cov(X_1 + X_2, Y) = Cov(X_1, Y) + Cov(X_2, Y)$

1.9 Basic concepts (Cont'd)

Conditional Expectation

• When X = x, the conditional expectation definition of Y is

$$E(Y \mid X = x) = \sum_{y} y p_{Y \mid X}(y \mid x)$$
 (Discrete)

$$E(Y \mid X = x) = \int_{y} y f_{Y \mid X}(y \mid x) dy$$
 (Continuous)

• Generally, the conditional expectation of function h(Y) is

$$E[h(Y) \mid X = x] = \sum_{y} h(y) p_{Y|X}(y \mid x) \quad \text{(Discrete)}$$

$$E[h(Y) \mid X = x] = \int_{y} h(y) f_{Y|X}(y \mid x) dy \quad \text{(Continuous)}$$

Theorem

$$E(Y) = E[E(Y \mid X)].$$

$$D(Y) = D[E(Y | X)] + E[D(Y | X)].$$

2.1 Extra Exercises 1

Expexctation calculation

• The CDF of R.V. X is:

$$F(x) = \begin{cases} \exp(x)/2, & x < 0 \\ 1/2, & 0 \le x < 1 \\ 1 - 1/2 \exp(-\frac{1}{2}(x - 1)), & x \ge 1 \end{cases}$$

Try to calculate E(X).

• R.V. X has $E(X) = Var(X) = \lambda$. If E(X - 1)(X - 2) = 1, what is λ ?

Expexciation and Probability

For non-negative R.V. X and $E(X) < +\infty$, try to prove: (Changeable of summation and integration)

- Discrete: $E(X) = \sum_{k=1}^{+\infty} P(X \ge k);$
- Continuous: $E(X) = \int_0^{+\infty} P(X > x) d_x$.

- **6.** Let *X* be a continuous random variable with probability density function $f(x) = 2x, 0 \le x \le 1$.
 - **a.** Find E(X).
 - **b.** Let $Y = X^2$. Find the probability mass function of Y and use it to find E(Y).
 - **c.** Use Theorem A in Section 4.1.1 to find $E(X^2)$ and compare to your answer in part (b).
 - d. Find Var(X) according to the definition of variance given in Section 4.2. Also find Var(X) by using Theorem B of Section 4.2.
- **15.** Suppose that two lotteries each have *n* possible numbers and the same payoff. In terms of expected gain, is it better to buy two tickets from one of the lotteries or one from each?
- 21. A random square has a side length that is a uniform [0, 1] random variable. Find the expected area of the square.
- **30.** Find E[1/(X+1)], where X is a Poisson random variable.



31. Let *X* be uniformly distributed on the interval [1, 2]. Find E(1/X). Is E(1/X) = 1/E(X)?

Supplementary Questions:

1. Suppose that the density function of a random variable is

$$f(x) = \begin{cases} 0, & x < 0 \\ e^{-x}, & x \ge 0 \end{cases}$$

Find the expectation of

(1)
$$Y = 2X$$

(2)
$$Y = e^{-2x}$$

2. Suppose that the joint density function of random variable (X,Y) is

$$f(x) = \begin{cases} 12y^2, & 0 < y < x < 1 \\ x, & \text{Otherwise} \end{cases}$$

Compute E(X), E(Y), E(XY), $E(X^2 + Y^2)$



2.3 Extra Exercises 2

Expexciation calculation for functions

• R.V. $X_i \stackrel{\text{iid}}{\sim} U(0,\theta)$, $i=1,\cdots,n$. Let $Y=\max\{X_1,X_2,\cdots,X_n\}$, $Z=\min\{X_1,X_2,\cdots,X_n\}$, try to calculate E(Y),E(Z). (Distribution of order statistics)

Conditional expexciation

Let X, Y are $\stackrel{\text{iid}}{\sim} \text{Exp}(\lambda)$. We define

$$Z = \begin{cases} 3X + 1, & X \ge Y \\ 6Y, & X < Y \end{cases}$$

Try to calculate E(Z). (Full expectation formula)



49. Two independent measurements, X and Y, are taken of a quantity μ . $E(X) = E(Y) = \mu$, but σ_X and σ_Y are unequal. The two measurements are combined by means of a weighted average to give

$$Z = \alpha X + (1 - \alpha)Y$$

where α is a scalar and $0 \le \alpha \le 1$.

- **a.** Show that $E(Z) = \mu$.
- **b.** Find α in terms of σ_X and σ_Y to minimize Var(Z).
- **c.** Under what circumstances is it better to use the average (X + Y)/2 than either *X* or *Y* alone?
- **50.** Suppose that X_i , where $i=1,\ldots,n$, are independent random variables with $E(X_i)=\mu$ and $\operatorname{Var}(X_i)=\sigma^2$. Let $\overline{X}=n^{-1}\sum_{i=1}^n X_i$. Show that $E(\overline{X})=\mu$ and $\operatorname{Var}(\overline{X})=\sigma^2/n$.

54. Let X, Y, and Z be uncorrelated random variables with variances σ_X^2 , σ_Y^2 , and σ_Z^2 , respectively. Let

$$U = Z + X$$
$$V = Z + Y$$

Find Cov(U, V) and ρ_{UV} .

- **55.** Let $T = \sum_{k=1}^{n} kX_k$, where the X_k are independent random variables with means μ and variances σ^2 . Find E(T) and Var(T).
- 60. Let Y have a density that is symmetric about zero, and let X = SY, where S is an independent random variable taking on the values +1 and −1 with probability ½ each. Show that Cov(X, Y) = 0, but that X and Y are not independent.

Supplementary Questions:

- 1. Suppose that X and Y are independent random variables. E(X) = 3, E(Y) = 1, D(X) = 4, D(Y) = 9. If Z = 5X 2Y + 15, compute E(Z), D(Z).
- 2. Suppose that X_i (i=1,2,3,4) are mutually independent to each other. (X_i) = 2i, $D(X_i) = 5 i$. If $Z = 2X_1 X_2 + 3X_3 0.5X_4$, compute E(Z) and D(Z).

P171: 54, 60 and the extra question

Extra: Assume that the density function for X is

$$f(x) = \frac{1}{2}e^{-|x|}, -\infty < x < \infty$$

- (1) Compute E(X) and D(X).
- (2) Are X and |X| independent or not? State your reason.
- (3) Are X and |X| correlated or not? State your reason



Supplementary Questions

1. Suppose that the joint density function of (X, Y) is

$$f(x,y) = \begin{cases} \frac{x+y}{8}, & 0 \le x \le 2, 0 \le y \le 2\\ 0, & \text{otherwise} \end{cases}$$
Compute $E(X)$, $E(Y)$, $Cov(X,Y)$, ρ_{XY} , $D(X+Y)$.

2. X and Y are independent random variables which both follow the normal distribution $N(\mu, \sigma^2)$. If $Z = \alpha X + \beta Y$, $W = \alpha X - \beta Y$, compute Cov(Z, W) and ρ_{ZW} .