

MATH 180A (Lecture A00)

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Today: Expectation of a function of a
random variable. Variance

Next: ASV 3.5

Week 6:

- Homework 4 due Friday, February

Expectation of continuous random variables

Example

Consider function $f(t) = \begin{cases} 0, & t \leq 1 \\ \frac{1}{t^2}, & t > 1 \end{cases}$

Is $f(t)$ a probability density?

$$f(t) \geq 0 \quad \int_{\mathbb{R}} f(t) dt =$$

Suppose that X is a random variable with PDF $f_X = f$.

What is $E(X)$?

$$E(X) = \int_{-\infty}^{+\infty} t f_X(t) dt =$$

Expectations of functions of random variables



Example $X \sim \text{Bin}(n, p)$ is the number of successes in n trials

$g(X) = \frac{X}{n}$ is the proportion of successful trials

$$g(X) \in \left\{ 0, \frac{1}{n}, \frac{2}{n}, \frac{3}{n}, \dots, \frac{n}{n} = 1 \right\},$$

$$E(g(X)) =$$

Proposition For a discrete random variable X

$$E(g(X)) =$$

Note, by definition $E(g(X)) = \sum_t t P(g(X)=t)$

Expectations of functions of random variables

Proposition For a continuous random variable X with density f_x

$$E(g(x)) =$$

Example Let U be a uniform random variable on $[a, b]$

Then

$$E[U^2] =$$

=

$$f_u(t) = \begin{cases} \frac{1}{b-a}, & t \in [a, b] \\ 0, & \text{otherwise} \end{cases}$$

Important class of functions: $g(x) = x^n$

discrete: $E(X^n) = \sum_t t^n P(X=t)$, continuous: $E(X^n) = \int_{-\infty}^{\infty} t^n f_x(t) dt$

Expectations of functions of random variables

Example (Car accident/insurance example)

An accident causes Y dollars of damage to your car, where insurance deductible is 500 dollars.

$Y \sim \text{Unif}([100, 1500])$. What is the expected amount you pay?

$X = \text{amount you pay} =$
(neither discrete
nor continuous)

$$E(X) = E(g(Y)) =$$

$$\min\{t, 1500\} = \begin{cases} t, & t \in [100, 500] \\ 500, & t \in [500, 1500] \end{cases}$$



Variance

Definition The variance of a random variable X is

- first compute $\mu = E(X)$, then compute $E(g(X))$ with $g(t) = (t - \mu)^2$
- if X is discrete, $\text{Var}(X) =$
- if X is continuous, $\text{Var}(X) =$
-
- The square root of the variance is called

Variance . Example

Example $X \sim \text{Ber}(p)$, $E(X) = p$

$$\text{Var}(X) =$$

=

=

Example $U \sim \text{Unif}[a, b]$, $E(U) = \frac{a+b}{2}$

$$\text{Var}(U) =$$

Alternative formula for variance

Proposition. Let X be a random variable. Then

$$\text{Var}(X) =$$

Proof (For continuous random variables) Let $\mu := E(X)$.

Then

$$\text{Var}(X) = E((X - \mu)^2) = \int_{\mathbb{R}} (t - \mu)^2 f_X(t) dt$$

=

=

=

Example $X \sim \text{Ber}(p)$, $E(X) = p$.

Variance

Variance is a measure of how "spread out from the mean" the distribution is.

Proposition Let X be a random variable with finite expectation $E(X) = \mu$. Then

Proof (\Leftarrow) Exercise

(\Rightarrow) (Assume X is discrete).

$$0 = \text{Var}(X) = \quad \Rightarrow \text{For all } t,$$

For all t , either \quad or \quad , so if
therefore,

Expectation and variance of $aX+b$

Let X be a random variable, and let $a, b \in \mathbb{R}$. Then

$$(i) E(aX+b) =$$

$$(ii) \text{Var}(aX+b) =$$

if $E(X)$ and $\text{Var}(X)$ exist

Proof (i) \leftarrow homework

$$(ii) \text{Var}(aX+b) =$$

$$=$$
$$=$$
$$=$$
$$=$$

Variance of geometric distribution

Let $X \sim \text{Geom}(p)$. We know that $E(X) = \frac{1}{p}$

$$E(X^2) = \sum_{k=1}^{\infty} k^2 P(X=k) = \sum_{k=1}^{\infty} k^2 \cdot p(1-p)^{k-1} =$$

=

$$l = k-1 =$$

=

"

Random variables. Summary

Discrete

Finite/countable set of possible values, $\sum_t P(X=t) = 1$

$$\text{PMF: } p_X(t) = P(X=t)$$

$$P(X \in B) = \sum_{t \in B} p_X(t)$$

CDF F_X is a step function

$$\text{Expectation: } E(X) = \sum_t t P(X=t)$$

$$E(g(X)) = \sum_t g(t) P(X=t)$$

Relation between CDF and PMF:
magnitude of jump of F_X at t is $P(X=t)$

Continuous

Uncountable set of possible values, $\forall t \in \mathbb{R} \quad P(X=t) = 0$

$$\text{PDF: } f_X: \mathbb{R} \rightarrow \mathbb{R}$$

$$P(X \in B) = \int_B f_X(t) dt$$

CDF F_X is a continuous function

$$\text{Expectation: } E(X) = \int_{\mathbb{R}} t f_X(t) dt$$

$$E(g(X)) = \int_{\mathbb{R}} g(t) f_X(t) dt$$

Relation between CDF and PDF:

$f_X(t) = F_X'(t)$ on the intervals where F_X is differentiable

Random variables. Summary

F_x is (i) nondecreasing, (ii) right-continuous

$$(iii) \lim_{t \rightarrow -\infty} F_x(t) = 0, \quad \lim_{t \rightarrow +\infty} F_x(t) = 1$$

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

$$E(aX + b) = aE(X) + b, \quad \text{Var}(aX + b) = a^2 \text{Var}(X)$$
