

# MATH 180A (Lecture A00)

[mathweb.ucsd.edu/~ynemish/teaching/180a](http://mathweb.ucsd.edu/~ynemish/teaching/180a)

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

Next: ASV 3.5

Week 6:

- Homework 4 due Friday, February 17

# 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?

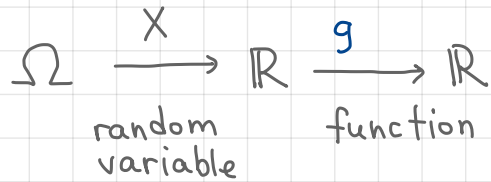
$$f(t) \geq 0 \quad \int_{\mathbb{R}} f(t) dt = \int_1^{+\infty} \frac{1}{t^2} dt = -\frac{1}{t} \Big|_1^{+\infty} = 1 \quad \checkmark$$

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

What is  $E(X)$ ?

$$E(X) = \int_{-\infty}^{+\infty} t f_X(t) dt = \int_1^{+\infty} t \cdot \frac{1}{t^2} dt = \int_1^{+\infty} \frac{1}{t} dt = \log t \Big|_1^{+\infty} = +\infty$$

# Expectations of functions of random variables



composition  $g(X) = g \circ X$

$g(X)$  is a new random variable

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

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

$$E(g(X)) = \sum_{k=0}^n \frac{k}{n} P(g(X) = \frac{k}{n}) = \sum_{k=0}^n \frac{k}{n} P(X=k) = \frac{E(X)}{n} = p$$

Proposition For a discrete random variable  $X$

$$E(g(X)) = \sum_s g(s) P(X=s)$$

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  $f_x$

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

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

Then

$$E[U^2] = \int_a^b t^2 \cdot f_u(t) dt$$

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

$$= \int_a^b t^2 \cdot \frac{1}{b-a} dt = \frac{1}{b-a} \cdot \left. \frac{t^3}{3} \right|_a^b = \frac{b^3 - a^3}{3(b-a)} = \frac{b^2 + ab + a^2}{3}$$

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

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

↔ moments ↔

# Expectations of functions of random variables

## Example (Car accident/insurance example)

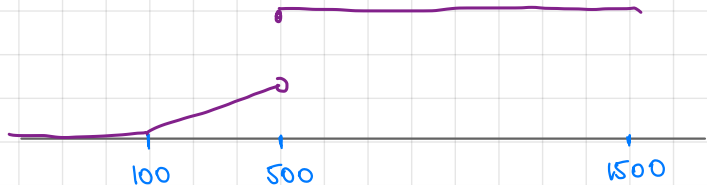
An accident causes  $Y$  dollars of damage to your car, where insurance deductible is 500 dollars.

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

$X = \text{amount you pay} = \min\{Y, 500\} = g(Y)$ ,  $g(t) = \min\{t, 500\}$   
(neither discrete nor continuous)

$$E(X) = E(g(Y)) = \int_{-\infty}^{+\infty} \min\{t, 500\} \cdot f_Y(t) dt = \int_{100}^{1500} \min\{t, 500\} \cdot \frac{1}{1400} dt$$
$$= \int_{100}^{500} \frac{t}{1400} dt + \int_{500}^{1500} \frac{500}{1400} dt = \frac{1}{1400} \left( \frac{500^2}{2} - \frac{100^2}{2} \right) + \frac{500}{1400} \cdot (1500 - 500) = 442.86$$

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



## Variance

Definition The variance of a random variable  $X$  is

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

- first compute  $\mu = E(X)$ , then compute  $E(g(X))$  with  $g(t) = (t - \mu)^2$
- if  $X$  is discrete,  $\text{Var}(X) = \sum_t (t - \mu)^2 P(X = t)$
- if  $X$  is continuous,  $\text{Var}(X) = \int_{-\infty}^{+\infty} (t - \mu)^2 f_X(t) dt$
- $\text{Var}(X) \geq 0$  (always)
- The square root of the variance is called standard deviation  $\sigma(X) = \sqrt{\text{Var}(X)}$

## Variance . Example

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

$$\text{Var}(X) = E((X-p)^2) = \sum_t (t-p)^2 P(X=t)$$

$$= (1-p)^2 \cdot p + (0-p)^2 \cdot (1-p) = p(1-p)^2 + p^2(1-p)$$

$$= p(1-p)(1-p+p) = p(1-p)$$

$$\sigma(X) = \sqrt{p(1-p)}$$

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

$$\text{Var}(U) = \int_{-\infty}^{+\infty} \left(t - \frac{a+b}{2}\right)^2 f_U(t) dt = \int_a^b \left(t - \frac{a+b}{2}\right)^2 \frac{1}{b-a} dt = \frac{(b-a)^2}{12}$$

$$\sigma(X) = \frac{|b-a|}{\sqrt{12}}$$

## Alternative formula for variance

Proposition. Let  $X$  be a random variable. Then

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

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

Then

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

$$= \int_{-\infty}^{+\infty} (t^2 - 2t\mu + \mu^2) f_X(t) dt$$

$$= \int_{-\infty}^{+\infty} t^2 f_X(t) dt - 2\mu \int_{-\infty}^{+\infty} t f_X(t) dt + \mu^2 \int_{-\infty}^{+\infty} f_X(t) dt$$

$$= E(X^2) - 2 \cdot \mu \cdot \mu + \mu^2 \cdot 1 = E(X^2) - \mu^2$$

Example  $X \sim \text{Ber}(p)$ ,  $E(X) = p$ ,  $E(X^2) = 1^2 \cdot p + 0^2 \cdot (1-p) = p$ ,  $\text{Var}(X) = p - p^2 = p(1-p)$



## 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,