

# 1.017/1.010 Class 8

## Expectation, Functions of a Random Variable

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Mean, variance of random variables

Expectation (population **mean**) of  $x$ ...  $E[x]$

For a **discrete**  $x$ :

$$E(x) = \bar{x} = \sum x_i p_x(x_i)$$

For a **continuous**  $x$ :

$$E(x) = \bar{x} = \int_{-\infty}^{+\infty} x f_x(x) dx$$

(Population) **variance** of  $x$ ...  $Var[x]$ :

$$Var(x) = E[(x - \bar{x})^2]$$

Derive these for uniform & triangular distributions

Functions of a random variable

$y = g(x)$   $x$  is a random variable with CDF  $F_x(x)$

$y$  is a random variable with CDF  $F_y(y)$  since it depends on  $x$

Derived distribution problems

Derive  $F_y(y)$  from  $F_x(x)$  using either of these options:

1. Analytical derivation

Apply definitions of  $y$ ,  $F_x(x)$ , and  $F_y(y)$ .

$$F_y(y) = P[y \leq y] = P[g(x) \leq y] = P[x = \{x \mid g(x) \leq y\}]$$

2. Stochastic simulation

Generate many realizations of  $x$ , compute  $y$  for each replicate, construct empirical  $F_y(y)$  from  $y$  replicates

Example (analytical derivation): Distribution of  $y = x^2$  for a uniformly distributed  $x$ :

Uniform distribution centered on 0:

$$y = g(x) = x^2$$

$$F_x(x) = \frac{x+1}{2} ; \quad f_x(x) = 1 ; \quad -1 \leq x \leq 1$$

$$F_y(y) = P[y \leq y] = P[g(x) \leq y] = P[x = \{x \mid g(x) \leq y\}]$$

$$F_y(y) = P[x = \{x \mid x^2 \leq y\}] = P[-y^{0.5} \leq x \leq y^{0.5}] = F_x(y^{0.5}) - F_x(-y^{0.5}) = y^{0.5} ; \quad 0 \leq y \leq 1$$

Uniform distribution centered on 0.5 (note need to split y interval into 2 parts):

$$y = g(x) = x^2$$

$$F_x(x) = (x+1)/3 ; \quad f_x(x) = 1/3 ; \quad -1 \leq x \leq 2$$

$$F_y(y) = P[y \leq y] = P[g(x) \leq y] = P[x = \{x \mid g(x) \leq y\}]$$

$$F_y(y) = P[x = \{x \mid x^2 \leq y\}] = P[-y^{0.5} \leq x \leq y^{0.5}] = F_x(y^{0.5}) - F_x(-y^{0.5}) = \frac{2y^{0.5}}{3} ; \quad 0 \leq y \leq 1$$

$$= P[x = \{x \mid x^2 \leq y\}] = P[-1 \leq x \leq y^{0.5}] = F_x(y^{0.5}) - F_x(-1) = (y^{0.5} + 1)/3 ; \quad 1 \leq y \leq 4$$

Mean and variance of  $y = g(x)$  :

$$E(y) = E[g(x)] = \sum_i g(x_i) p_x(x_i)$$

$$E(y) = E[g(y)] = \int_{-\infty}^{+\infty} g(x) f_x(x) dx$$

$$Var(y) = Var[g(y)] = E\{g(x) - \bar{g(x)}]^2\}$$



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