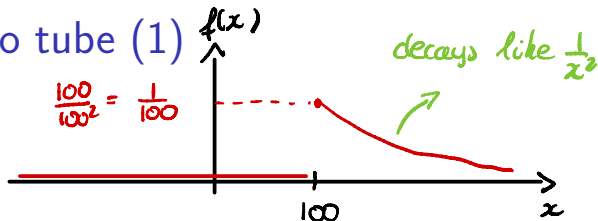


Example: radio tube (1)



Situation:

- X = lifetime of a radio tube
- Density of X :

$$f(x) = \frac{100}{x^2} \mathbf{1}_{(100, \infty)}(x)$$

- We have 5 tubes in a set

Question: Probability that 2 of the 5 tubes have to be replaced within the first 150h of operation

First step: For the rv X , compute

$$P(X \leq 150) = P(X \in (-\infty, 150])$$

$$= 0 \text{ if } x \leq 100$$

$$= \int_{-\infty}^{150} f(x) dx$$

$$= \int_{100}^{150} \frac{100}{x^2} dx$$

$$= 100 \int_{100}^{150} \frac{1}{x^2} dx$$

$$= 100 \left. -\frac{1}{x} \right|_{100}^{150}$$

$$= 100 \left(\frac{1}{100} - \frac{1}{150} \right) = 1 - \frac{2}{3} = \frac{1}{3}$$

Step 2 : Take the 5 components into account

Define , for $i = 1, \dots, 5$,

$$Y_i = \begin{cases} 1 & \text{if } i\text{-th component fails} \\ & \text{in the 1st 150h} \\ 0 & \text{otherwise} \end{cases}$$

Then Y_i is discrete. In fact

$$Y_i \sim B(p = \frac{1}{3})$$

Y_i 's are II

Rv of interest

$Z = \#$ components failing in the
1st 150 h

Then $Z = \sum_{i=1}^5 Y_i \sim \text{Bin}(n=5, p=\frac{1}{3})$

We wish to compute

$$P(Z = 2) = \binom{5}{2} \left(\frac{1}{3}\right)^2 \left(\frac{2}{3}\right)^3$$

$$\approx 33\%$$

Example: radio tube (2)

Family of events: We define

- $X_i =$ lifetime of tube i
- $E_i =$ "tube i has to be replaced within the first 150h of operation"

Probability of E_i :

$$\begin{aligned}\mathbf{P}(E_i) &= \mathbf{P}(X_i \leq 150) \\ &= \int_{-\infty}^{150} f(x) dx \\ &= 100 \int_{100}^{150} \frac{dx}{x^2}\end{aligned}$$

Thus

$$\mathbf{P}(E_i) = \frac{1}{3}$$

Example: radio tube (3)

Model for the set of tubes: Define

$$Z_i = \mathbf{1}_{E_i}, \quad Z = \sum_{i=1}^5 Z_i$$

Then

$$Z \sim \text{Bin}\left(5, \frac{1}{3}\right)$$

and we look for

$$\mathbf{P}(Z = 2)$$

Conclusion:

$$\mathbf{P}(Z = 2) = \binom{5}{2} \left(\frac{1}{3}\right)^2 \left(\frac{2}{3}\right)^3 \simeq 33\%$$

Outline

- 1 Introduction
- 2 Expectation and variance of continuous random variables**
- 3 The uniform random variable
- 4 Normal random variables
- 5 Exponential random variables
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Discrete case

$$E[X] = \sum_i x_i p(x_i)$$

Discrete \rightarrow continuous

pmf $p \rightarrow$ density f

$\sum \rightarrow \int$

Thus

$$E[X] = \int x f(x) dx$$

General definition

Definition 2.

Let

- \mathbf{P} a probability on a sample space S
- $X : S \rightarrow \mathbb{R}$ a continuous random variable
- $f =$ density of X

Then the expected value of X is defined by

$$\mathbf{E}[X] = \int_{\mathbb{R}} x f(x) dx$$

Heuristics for the definition

Recall the discrete case:

$$\mathbf{E}[X] = \sum_{i \geq 1} x_i \mathbf{P}(X = x_i)$$

Continuous case analog: We have

$$f(x) dx \simeq \mathbf{P}(x \leq X \leq x + dx)$$

Thus

$$\begin{aligned} \mathbf{E}[X] &\simeq \sum x_i \mathbf{P}(x_i \leq X \leq x_i + dx) \\ &\simeq \int_{\mathbb{R}} x f(x) dx \end{aligned}$$

Simple example (1)

Density of X : Consider X with density

$$f(x) = 2x \mathbf{1}_{[0,1]}(x)$$

Example $f(x) = 2x \mathbb{1}_{[0,1]}(x)$

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

$$= \int_0^1 x \times 2x dx$$

$$= 2 \int_0^1 x^2 dx$$

$$= 2 \left. \frac{x^3}{3} \right|_0^1$$

$$= \frac{2}{3}$$

Security check If f is a density,
then

(i) $f(x) \geq 0$ for all $x \in \mathbb{R}$

(ii) $\int_{-\infty}^{\infty} f(x) dx = 1$ (discrete: $\sum p(x) = 1$)

Here, with $f(x) = 2x \mathbb{1}_{(0,1)}(x)$,

$$\int_{-\infty}^{\infty} f(x) dx = \int_0^1 f(x) dx$$
$$= \int_0^1 (2x) dx = x^2 \Big|_0^1 = 1$$

OK!

Simple example (2)

Recall: We consider X with density

$$f(x) = 2x \mathbf{1}_{[0,1]}(x)$$

Expected value:

$$\begin{aligned} \mathbf{E}[X] &= \int_{\mathbb{R}} x f(x) dx \\ &= \int_0^1 2x^2 dx \\ &= \frac{2}{3} \end{aligned}$$

Expression for $\mathbf{E}[X]$ when $X \geq 0$

Proposition 3.

Let

- X continuous random variable
- f density of X

Hypothesis:

$$f(x) = 0 \text{ if } x < 0$$

$$X \geq 0$$

Then

$$\mathbf{E}[X] = \int_0^{\infty} \overbrace{\mathbf{P}(X > y)}^{\text{tail of } X} dy \quad (1)$$

Proof

Expression for the rhs:

$$\int_0^{\infty} \mathbf{P}(X > y) dy = \int_0^{\infty} \left(\int_y^{\infty} f(x) dx \right) dy$$

Apply Fubini: Invert the order of integration

$$\begin{aligned} \int_0^{\infty} \mathbf{P}(X > y) dy &= \int_0^{\infty} \left(\int_0^x dy \right) f(x) dx \\ &= \int_0^{\infty} x f(x) dx \\ &= \mathbf{E}[X] \end{aligned}$$

Definition of $\mathbf{E}[g(X)]$

$$\text{discrete} : \mathbf{E}[g(X)] = \sum_i g(x_i) p(x_i)$$

Proposition 4.

Let

- X continuous random variable
- f density of X
- g real valued function

Then

$$\mathbf{E}[g(X)] = \int_{\mathbb{R}} g(x) f(x) dx \quad (2)$$

Simple example – Ctd (1)

Density of X : Consider X with density

$$f(x) = 2x \mathbf{1}_{[0,1]}(x)$$

Question: Compute

$$\mathbf{E}[X^3]$$

Example - ctd

$$f(x) = 2x \mathbb{1}_{(0,1)}(x)$$

Then

$$\boxed{E[X^3]} = \int_{-\infty}^{\infty} x^3 f(x) dx$$

$$= \int_0^1 x^3 (2x) dx$$

$$= 2 \int_0^1 x^4 dx$$

$$= 2 \left. \frac{x^5}{5} \right|_0^1$$

$$\boxed{= \frac{2}{5}}$$

Simple example – Ctd (2)

Recall: We consider X with density

$$f(x) = 2x \mathbf{1}_{[0,1]}(x)$$

Expected value for $g(x) = x^3$:

$$\begin{aligned} \mathbf{E}[X^3] &= \int_{\mathbb{R}} x^3 f(x) dx \\ &= \int_0^1 2x^4 dx \\ &= \frac{2}{5} \end{aligned}$$

Proof of Proposition 4

Hypothesis:

We assume $X \geq 0$ and $g(X) \geq 0$ for the proof

Expression with (1):

$$\begin{aligned}\mathbf{E}[g(X)] &= \int_0^{\infty} \mathbf{P}(g(X) > y) dy \\ &= \int_0^{\infty} \left(\int_{\{x; g(x) > y\}} f(x) dx \right) dy\end{aligned}$$

Apply Fubini: Invert the order of integration

$$\begin{aligned}\mathbf{E}[g(X)] &= \int_0^{\infty} \left(\int_0^{g(x)} dy \right) f(x) dx \\ &= \int_0^{\infty} g(x) f(x) dx\end{aligned}$$

Expectation and linear transformations

Proposition 5.

Let

- X continuous random variable
- f density of X
- $a, b \in \mathbb{R}$ constants

Then

$$\mathbf{E}[aX + b] = a \mathbf{E}[X] + b$$

Proof

Application of relation (2):

$$\begin{aligned}\mathbf{E}[aX + b] &= \int_{\mathbb{R}} (ax + b) f(x) dx \\ &= a \int_{\mathbb{R}} x f(x) dx + b \int_{\mathbb{R}} f(x) dx \\ &= a\mathbf{E}[X] + b\end{aligned}$$

Definition of variance

Definition 6.

Let

- X continuous random variable
- f density of X
- $\mu = \mathbf{E}[X]$

Then we define $\mathbf{Var}(X)$ by

$$\mathbf{Var}(X) = \mathbf{E}[(X - \mu)^2]$$

Alternative expression for the variance

Proposition 7.

Let

- X continuous random variable
- f density of X
- $\mu = \mathbf{E}[X]$

Then $\mathbf{Var}(X)$ can be written as

$$\mathbf{Var}(X) = \mathbf{E}[X^2] - \mu^2 = \mathbf{E}[X^2] - (\mathbf{E}[X])^2$$

Example - ctd $f(x) = 2x \mathbb{1}_{(0,1)}(x)$
Then

$$\begin{aligned} E[X^2] &= \int_0^1 x^2 (2x) dx \\ &= 2 \int_0^1 x^3 dx = 2 \times \frac{x^4}{4} \Big|_0^1 \\ &= \frac{2}{4} = \frac{1}{2} \end{aligned}$$

$$\text{Var}(X) = \overbrace{E[X^2]}^{\frac{1}{2}} - \left(\overbrace{E[X]}^{\frac{2}{3}} \right)^2$$

$$= \frac{1}{2} - \left(\frac{2}{3} \right)^2 = \frac{1}{18}$$

Security check: verify that $\text{Var}(X) > 0$

Simple example – Ctd

Density of X : Consider X with density

$$f(x) = 2x \mathbf{1}_{[0,1]}(x)$$

Expected value for $g(x) = x^2$:

$$\mathbf{E}[X^2] = \int_{\mathbb{R}} x^2 f(x) dx = \int_0^1 2x^3 dx = \frac{1}{2}$$

Variance of X :

$$\mathbf{Var}(X) = \mathbf{E}[X^2] - (\mathbf{E}[X])^2 = \frac{1}{2} - \left(\frac{2}{3}\right)^2 = \frac{1}{18}$$

Variance and linear transformations

Proposition 8.

Let

- X continuous random variable
- f density of X
- $a, b \in \mathbb{R}$ constants

Then

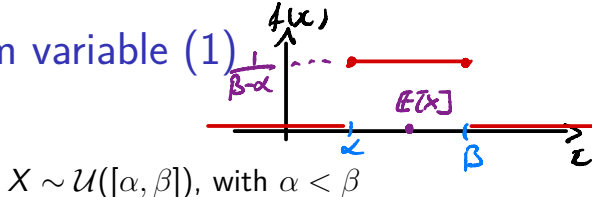
$$\mathbf{Var}(aX + b) = a^2 \mathbf{Var}(X)$$

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Uniform random variable (1)

Notation:



State space:

$$X \in [\alpha, \beta]$$

Density:

$f(x)$ is constant (uniform) on $[\alpha, \beta]$

$$f(x) = \frac{1}{\beta - \alpha} \mathbf{1}_{[\alpha, \beta]}(x)$$

Expected value and variance:

$$\mathbf{E}[X] = \frac{\alpha + \beta}{2}, \quad \mathbf{Var}(X) = \frac{(\beta - \alpha)^2}{12}$$