

More review.

Example. We say that Y is exponential w/ parameter $\tau > 0$ if it has the following pdf

$$f_Y(y) = \frac{1}{\tau} e^{-\frac{y}{\tau}} \mathbb{1}_{[0, \infty)}(y)$$

Alternative parameterization : $\lambda = \frac{1}{\tau}$

Expectations and Standard Deviations.

In the discrete case: $E[Y] = \sum_{y \in S_Y} y P_Y(y)$

Def'n. For a continuous r.v. Y w/ pdf f_Y , we define the expectation $E[Y]$ as

$$E[Y] := \int_{-\infty}^{+\infty} y \cdot f_Y(y) dy$$

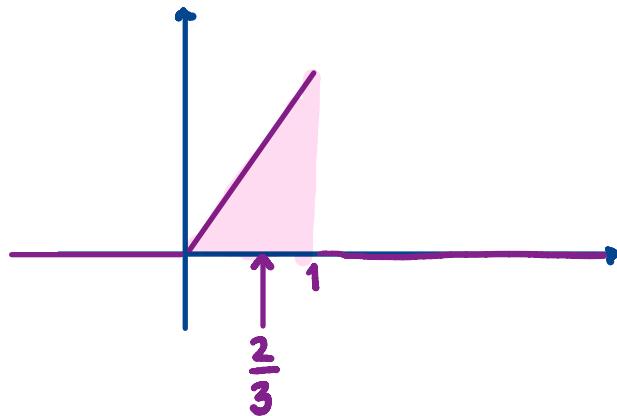
If the integral exists.

Problem 3.3. Consider a continuous random variable Y whose probability density function is given by

$$f_Y(y) = 2y \mathbf{1}_{[0,1]}(y)$$

What is the expected value of this random variable?

$$\rightarrow: \mathbb{E}[Y] = \int_0^1 y(2y) dy = 2 \int_0^1 y^2 dy = 2 \frac{y^3}{3} \Big|_{y=0}^1 - \frac{2}{3} \quad \square$$



Example .

$$Y \sim U(a, b)$$

$$\mathbb{E}[Y] = \frac{a+b}{2}$$

$$\text{Var}[Y] = ?$$

$$\begin{aligned}\text{Var}[Y] &= \mathbb{E}[(Y - \mu_Y)^2] \\ &= \mathbb{E}[Y^2] - \mu_Y^2\end{aligned}$$

$$Y - a \sim U(0, b-a)$$

$$U := \frac{Y-a}{b-a} \sim U(0,1)$$

$$Y = a + (b-a) \cdot U$$

$$\text{Var}[Y] = \text{Var}[a + (b-a) \cdot U] = (b-a)^2 \cdot \text{Var}[U]$$

deterministic

$$\text{Var}[U] = \mathbb{E}[U^2] - (\mathbb{E}[U])^2$$

$$\mathbb{E}[U^2] = \int_0^1 u^2 f_U(u) du = \int_0^1 u^2 du = \frac{u^3}{3} \Big|_{u=0}^1 = \frac{1}{3}$$

pdf of $U \sim U(0,1)$

$$\text{Var}[U] = \frac{1}{3} - \left(\frac{1}{2}\right)^2 = \frac{1}{12}$$

$$\text{Var}[Y] = \frac{(b-a)^2}{12}$$

□

Example. $Y \sim N(\mu, \sigma^2)$

$$\mathbb{E}[Y] \quad \text{SD}[Y]$$

Example. $Y \sim E(\tau)$, $\tau < \infty$ Y is exponential w/ parameter τ

$$\mathbb{E}[Y] = \int_{-\infty}^{+\infty} y f_Y(y) dy$$

$$= \int_{-\infty}^{+\infty} y \cdot \frac{1}{\tau} e^{-\frac{y}{\tau}} \mathbb{1}_{[0, \infty)}(y) dy$$

$$= \int_0^{+\infty} \left(\frac{y}{\tau}\right) e^{-\frac{(y)}{\tau}} dy$$

$$u = \frac{y}{\tau} \quad du = \frac{dy}{\tau} \quad dy = \tau \cdot du$$

$$= \tau \int_0^{+\infty} u e^{-u} du$$

$$u = u$$

$$dv = e^{-u} du$$

$$= \tau \left(-ue^{-u} \Big|_{u=0}^{+\infty} + \int_0^{+\infty} (-e^{-u}) du \right)$$

$$du = du$$

$$v = -e^{-u}$$

$$+ \int_0^{+\infty} (te^{-u}) du = \tau$$

$$-e^{-u} \Big|_{u=0}^{+\infty} = -(0 - 1) = 1$$

Def'n. For a r.v. Y w/ pdf f_Y and $k = 1, 2, \dots$, we define the k^{th} (raw) moment μ_{kY} by

$$\mu_{kY} = \mathbb{E}[Y^k] = \int_{-\infty}^{+\infty} y^k f_Y(y) dy ;$$

$$\boxed{\mu_1 = \mathbb{E}[Y]}$$

the k^{th} central moment μ_k^c is defined by

$$\mu_k^c = \mathbb{E} \left[(Y - \mu_Y)^k \right] = \int_{-\infty}^{\infty} (y - \mu_Y)^k f_Y(y) dy$$

$$\text{w/ } \mu_Y = \mu_1 ;$$

the skewness is

$$\mathbb{E} \left[\left(\frac{Y - \mu_Y}{\sigma_Y} \right)^3 \right] \quad \text{w/ } \sigma_Y^2 = \mathbb{E}[(Y - \mu_Y)^2] ;$$

the kurtosis is

$$\mathbb{E} \left[\left(\frac{Y - \mu_Y}{\sigma_Y} \right)^4 \right] = \frac{\mathbb{E}[(Y - \mu_Y)^4]}{\sigma_Y^4}$$

Cumulative Distribution Function.

Defn. The cumulative distribution function (cdf) of a r.v. Y is a function

$$F_Y : \mathbb{R} \longrightarrow [0, 1]$$

defined as

$$F_Y(y) = \mathbb{P}[Y \leq y] \quad \text{for all } y \in \mathbb{R}$$