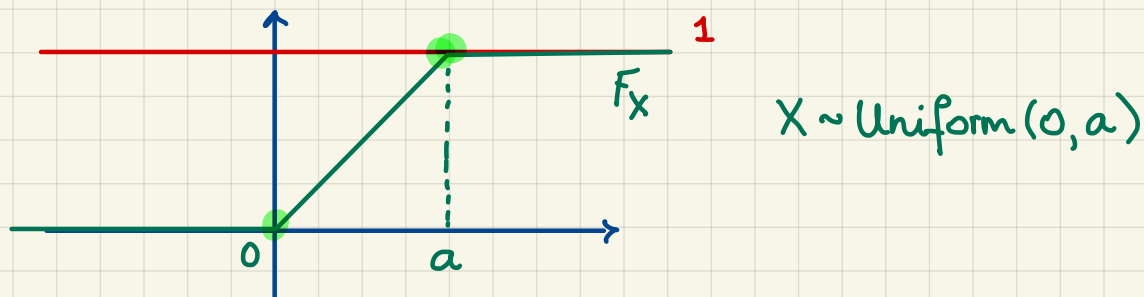


M358K: Applied Statistics.

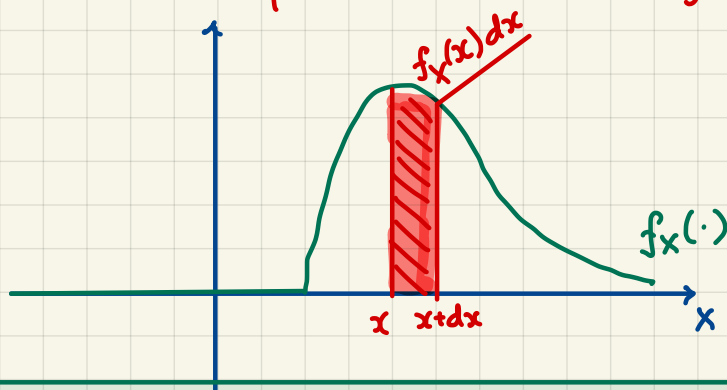
Continuous Random Variables [Review].

Def'n. A random variable X is said to be **continuous** if its cumulative distribution function F_X is:

- (i) continuous everywhere;
- (ii) differentiable everywhere except @ at most countably many points.



Def'n. Any function $f_X : \mathbb{R} \rightarrow [0, +\infty)$ such that $f_X(x) = F_X'(x)$ for all x where the derivative exists is called the **probability density function (pdf)** of X .



$$\textcircled{Q}: \mathbb{P}[a < X \leq b] = \int_a^b f_X(x) dx = F_X(b) - F_X(a) \textcircled{\smiley}$$

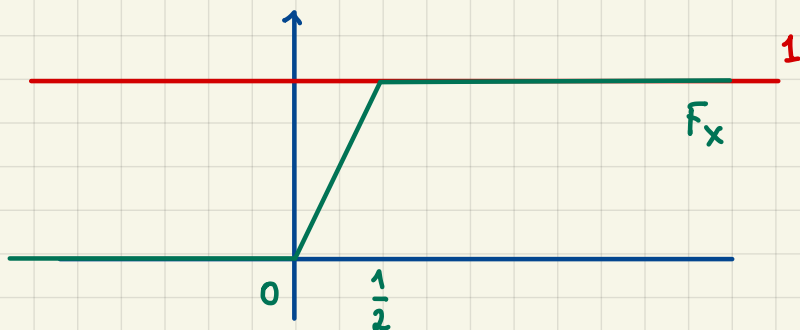
$$\textcircled{Q}: X \text{ is continuous} \Rightarrow \mathbb{P}[X=x] = 0$$

$$\textcircled{Q}: \int_{-\infty}^{+\infty} f_X(x) dx = 1$$

Q: Is it possible that $f_X(x) > 1$ for some x ? Yes.

Example.

$$X \sim U(0, \frac{1}{2})$$



$$F_X(x) = \begin{cases} 0 & \text{for } x < 0 \\ 2x & \text{for } 0 \leq x \leq \frac{1}{2} \\ 1 & \text{for } x > \frac{1}{2} \end{cases}$$

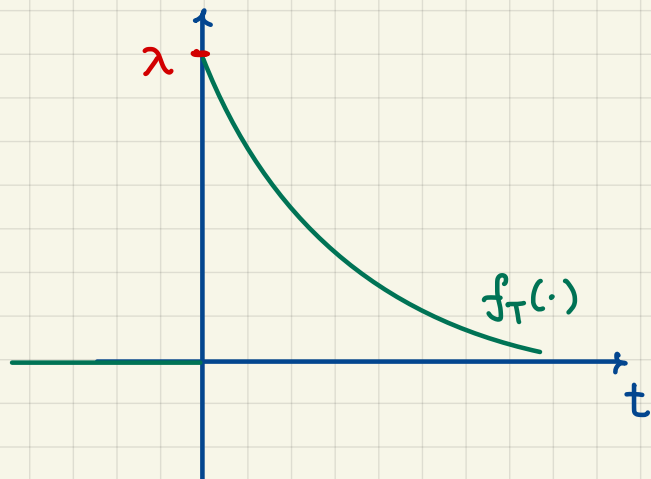
$$f_X(x) = \begin{cases} 0 & \text{for } x < 0 \\ 2 & \text{for } 0 \leq x \leq \frac{1}{2} \\ 0 & \text{for } x > \frac{1}{2} \end{cases}$$

Example. Exponential Distribution. $T \sim \text{Exp}(\lambda)$

w/ a positive parameter λ

Its pdf is:

$$f_T(t) = \begin{cases} \lambda e^{-\lambda t} & t > 0 \\ 0 & \text{otherwise} \end{cases}$$



Note: If $\lambda = 1.77$,
then $f_T(t) \approx 1.77$
for $t \approx 0$

Expected Value.

Def'n. For a discrete r.v. X , its **expected value** (expectation/mean) is given by:

$$\mathbb{E}[X] := \sum_x p_X(x) \cdot x$$

when the sum exists

For a continuous r.v. X , its **expected value** (expectation/mean) is given by:

$$\mathbb{E}[X] := \int_{-\infty}^{+\infty} x \cdot f_X(x) dx$$

when the integral exists

Example. $T \sim \text{Exp}(\lambda)$

$$\Rightarrow \mathbb{E}[T] = \frac{1}{\lambda}$$

Def'n. For any r.v. X , its **variance** is defined as:

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

if it exists

Note:

Set $\mu_X := \mathbb{E}[X]$

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

$$= \mathbb{E}[X^2 - 2 \cdot \mu_X \cdot X + \mu_X^2]$$

$$= \mathbb{E}[X^2] - 2\mu_X \mathbb{E}[X] + \mu_X^2$$

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

Def'n. The **standard deviation** of the r.v. X is:

$$\text{SD}[X] = \sqrt{\text{Var}[X]}$$