

From Discrete to Continuous

- So far, all random variables we saw were *discrete*
 - Have finite or countably infinite values (e.g., integers)
 - Usually, values are binary or represent a count
- Now it's time for *continuous* random variables
 - Have (uncountably) infinite values (e.g., real numbers)
 - Usually represent measurements (arbitrary precision)
 - Height (centimeters), Weight (lbs.), Time (seconds), etc.
- Difference between how many and how much
- Generally, it means replace $\sum_{x=a}^b f(x)$ with $\int_a^b f(x)dx$

Continuous Random Variables

- X is a **Continuous Random Variable** if there is function $f(x) \geq 0$ for $-\infty \leq x \leq \infty$, such that:

$$P(a \leq X \leq b) = \int_a^b f(x)dx$$

- f is a Probability Density Function (PDF) if:

$$P(-\infty < X < \infty) = \int_{-\infty}^{\infty} f(x)dx = 1$$

Probability Density Functions

- Say f is a **Probability Density Function** (PDF)

$$P(-\infty < X < \infty) = \int_{-\infty}^{\infty} f(x)dx = 1$$

- $f(x)$ is not a probability, it is probability/units of X
- Not meaningful without some subinterval over X

$$P(X = a) = \int_a^a f(x)dx = 0$$

- Contrast with Probability Mass Function (PMF) in discrete case: $p(a) = P(X = a)$

where $\sum_{i=1}^{\infty} p(x_i) = 1$ for X taking on values x_1, x_2, x_3, \dots

Cumulative Distribution Functions

- For a continuous random variable X , the **Cumulative Distribution Function** (CDF) is:

$$F(a) = P(X < a) = P(X \leq a) = \int_{-\infty}^a f(x)dx$$

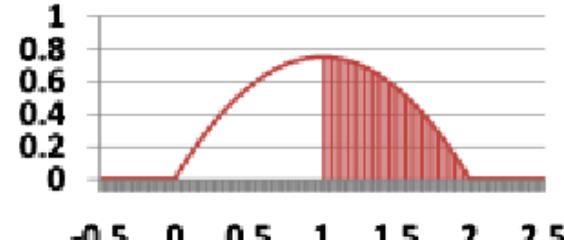
- Density f is derivative of CDF F : $f(a) = \frac{d}{da} F(a)$
- For continuous f and small ε :

$$P\left(a - \frac{\varepsilon}{2} \leq X \leq a + \frac{\varepsilon}{2}\right) = \int_{a - \varepsilon/2}^{a + \varepsilon/2} f(x)dx \approx \varepsilon f(a)$$

Simple Example

- X is continuous random variable (CRV) with PDF:

$$f(x) = \begin{cases} C(4x - 2x^2) & \text{when } 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$



- What is C?

$$\int_0^2 C(4x - 2x^2) dx = 1 \Rightarrow C \left[2x^2 - \frac{2x^3}{3} \right]_0^2 = 1$$

$$C \left[\left(8 - \frac{16}{3} \right) - 0 \right] = 1 \Rightarrow C \frac{8}{3} = 1 \Rightarrow C = \frac{3}{8}$$

- What is $P(X > 1)$?

$$\int_1^\infty f(x) dx = \int_1^2 \frac{3}{8}(4x - 2x^2) dx = \frac{3}{8} \left[2x^2 - \frac{2x^3}{3} \right]_1^2 = \frac{3}{8} \left[\left(8 - \frac{16}{3} \right) - \left(2 - \frac{2}{3} \right) \right] = \frac{1}{2}$$

Disk Crashes

- $X = \text{hours before your disk crashes}$

$$f(x) = \begin{cases} \lambda e^{-x/100} & x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

- First, determine λ to have actual PDF

- Good integral to know: $\int e^u du = e^u$

$$1 = \int \lambda e^{-x/100} dx = -100\lambda \int \frac{-1}{100} e^{-x/100} dx = -100\lambda e^{-x/100} \Big|_0^\infty = 100\lambda \Rightarrow \lambda = \frac{1}{100}$$

- What is $P(50 < X < 150)$?

$$F(150) - F(50) = \int_{50}^{150} \frac{1}{100} e^{-x/100} dx = -e^{-x/100} \Big|_{50}^{150} = -e^{-3/2} + e^{-1/2} \approx 0.383$$

- What is $P(X < 10)$?

$$F(10) = \int_0^{10} \frac{1}{100} e^{-x/100} dx = -e^{-x/100} \Big|_0^{10} = -e^{-1/10} + 1 \approx 0.095$$

Expectation and Variance

For discrete RV X :

$$E[X] = \sum_x x p(x)$$

$$E[g(X)] = \sum_x g(x) p(x)$$

$$E[X^n] = \sum_x x^n p(x)$$

For continuous RV X :

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

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

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

For both discrete and continuous RVs:

$$E[aX + b] = aE[X] + b$$

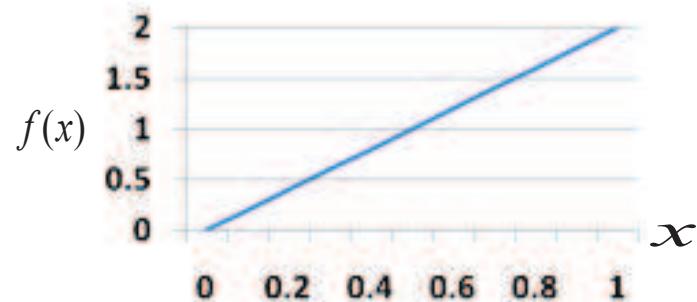
$$\text{Var}(X) = E[(X - \mu)^2] = E[X^2] - (E[X])^2$$

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

Linearly Increasing Density

- X is a continuous random variable with PDF:

$$f(x) = \begin{cases} 2x & 0 \leq x \leq 1 \\ 0 & \text{otherwise} \end{cases}$$



- What is $E[X]$?

$$E[X] = \int_{-\infty}^{\infty} x f(x) dx = \int_0^1 2x^2 dx = \frac{2}{3} x^3 \Big|_0^1 = \frac{2}{3}$$

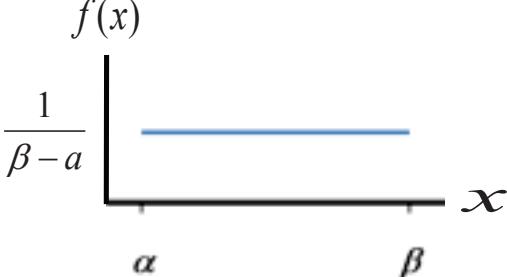
- What is $\text{Var}(X)$?

$$E[X^2] = \int_{-\infty}^{\infty} x^2 f(x) dx = \int_0^1 2x^3 dx = \frac{1}{2} x^4 \Big|_0^1 = \frac{1}{2}$$

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

Uniform Random Variable

- X is a **Uniform Random Variable**: $X \sim \text{Uni}(\alpha, \beta)$
 - Probability Density Function (PDF):

$$f(x) = \begin{cases} \frac{1}{\beta-\alpha} & \alpha < x < \beta \\ 0 & \text{otherwise} \end{cases}$$


- $P(a \leq x \leq b) = \int_a^b f(x)dx = \frac{b-a}{\beta-\alpha}$
- $E[X] = \int_{-\infty}^{\infty} xf(x)dx = \int_{-\infty}^{\infty} \frac{x}{\beta-\alpha} dx = \frac{\beta^2 - \alpha^2}{2(\beta-\alpha)} = \frac{\alpha + \beta}{2}$
- $Var(X) = \frac{(\beta-\alpha)^2}{12}$