

Continuous Random Variables

So far, all random variables we have seen have been *discrete*. In all the cases we have seen in CS109 this meant that our RVs could only take on integer values. Now it's time for *continuous* random variables which can take on values in the real number domain. They usually represent measurements with arbitrary precision (eg height, weight, time).

Continuous Random Variables

Probability Density Functions

X is a Continuous Random Variable if there is a Probability Density Function (PDF) $f(x)$ for $-\infty \leq x \leq \infty$ such that:

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

The following properties must also hold. These preserve the axiom that $P(a \leq X \leq b)$ is a probability:

$$0 \leq P(a \leq X \leq b) \leq 1$$

$$P(-\infty < X < \infty) = 1$$

A common misconception is to think of $f(x)$ as a probability. It is instead what we call a probability density. It represents probability/unit of X . Generally this is not particularly meaningful without either taking the interval over X or comparing it to another probability density. Of special note, the probability that a continuous random variable takes on a specific value (to infinite precision) is 0.

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

That is pretty different than in the discrete world where we often talked about the probability of a random variable taking on a particular value.

Cumulative Distribution Function

For a continuous random variable X the Cumulative Distribution Function, written $F(a)$ or as (CDF) is:

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

Example 1

Let X be a continuous random variable (CRV) with PDF:

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

In this function, C is a constant. What value is C ? Since we know that the PDF must sum to 1:

$$\int_0^2 C(4x - 2x^2)dx = 1$$

$$C \left(2x^2 - \frac{2x^3}{3} \right) \Big|_0^2 = 1$$

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

And if you solve the equation for C you find that $C = 3/8$.

What is $P(X > 1)$

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

Example 2

Let X be a random variable which represents the number of days of use before your disk crashes with PDF:

$$f(x) = \begin{cases} \lambda 3^{x/100} & \text{when } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

First, determine λ . Recall that $\int e^u du = e^u$:

$$\begin{aligned} \int \lambda 3^{x/100} dx &= 1 \Rightarrow -100\lambda \int \frac{-1}{100} 3^{x/100} dx = 1 \\ -100\lambda 3^{-x/100} \Big|_0^\infty &= 1 \Rightarrow 100\lambda = 1 \Rightarrow \lambda = \frac{1}{100} \end{aligned}$$

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

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

Expectation and Variance

For continuous RV X :

$$\begin{aligned} E[X] &= \int_{-\infty}^{\infty} x f(x) dx \\ E[g(X)] &= \int_{-\infty}^{\infty} g(x) f(x) dx \\ E[X^n] &= \int_{-\infty}^{\infty} x^n f(x) dx \end{aligned}$$

For both continuous and discrete RVs:

$$\begin{aligned} E[aX + b] &= aE[X] + b \\ \text{Var}(X) &= E[(X - \mu)^2] = E[X^2] - (E[X])^2 \end{aligned}$$

Uniform Random Variable

X is a Uniform Random Variable $X \sim \text{Uni}(\alpha, \beta)$ if:

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

The key properties of this RV are:

$$\begin{aligned} P(a \leq x \leq b) &= \int_a^b f(x) dx = \frac{b - a}{\beta - \alpha} \quad (\text{for } \alpha \leq a \leq b \leq \beta) \\ E[X] &= \int_{-\infty}^{\infty} x f(x) dx = \int_{\alpha}^{\beta} \frac{x}{\beta - \alpha} dx = \frac{x^2}{2(\beta - \alpha)} \Big|_{\alpha}^{\beta} = \frac{\alpha + \beta}{2} \\ \text{Var}(X) &= \frac{(\beta - \alpha)^2}{12} \end{aligned}$$