DSC 215 - PROBABILITY AND STATISTICS FOR DATA SCIENCE

RANDOMVARIABLES



COMPUTER SCIENCE & ENGINEERING

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Random Variables

- In the context of statistics, random variables (rv) allow us to apply a mathematical framework for better understanding and predicting outcomes in the real world.
- Mathematically, **a random variable** is a function from Ω to another set, such as the real line \mathbb{R} , or the natural numbers (positive integers) \mathbb{N} .
- If the range is, say, \mathbb{R} , we have the random variable $X : \Omega \to \mathbb{R}$, which means that for $\omega \in \Omega$, we may write $X(\omega) = x \in \mathbb{R}$.

Example: Random Variables

Example:

- Consider Ω to be the set of all length 3 sequences of Heads and Tails so that, for example

$$(H, H, T) \in \Omega$$

$$\underline{\qquad}$$

• Suppose the random variable X is the function that counts how many heads are in the sequence. So, in our example

$$\omega = (H, H, T) \implies X(\omega) = 2$$

Discrete VS Continuous Random Variables

- **Discrete random variables:** In our previous example $X: \Omega \to \{0,1,2,3\}$ so we call it a **discrete random variable** (because the range is a discrete set). Note that the discrete set can be infinite, like \mathbb{N} .
- We can expand our notation to more conveniently write the probability that there are k heads in our sequence as

$$\mathbb{P}(X=k) := \mathbb{P}(\{\omega : X(\omega)=k\}).$$

• Continuous random variables: If, for example $X:\Omega\to\mathbb{R}$, we call it a continuous random variable. Here we can introduce the convenient notation

$$\mathbb{P}(a \le X \le b) := \mathbb{P}(\{\omega : a \le X(\omega) \le b\}).$$

Probability Mass Functions (Discrete Random Variables)

• For a discrete random variable the probability mass function (PMF) is the function

$$p_X: \mathbb{R} \to [0,1]$$
 with $p_X(a) = \mathbb{P}(X=a)$.

- It gives the probability that a discrete random variable is exactly equal to some value.
- PMFs satisfy
 - $p_X(x) \ge 0$ (all probabilities are non-negative)
 - $\sum_{x} p_X(x) = 1$ (because probabilities of all disjoint events sum to 1)

Probability Mass Functions

Example:

• Let X be a random variable associated with a coin-tossing experiment, so $\Omega = \{H, T\}$, and suppose that X(H) = 1 and X(T) = 0.

• If the coin is fair:
$$p_X(x) = \begin{cases} 1/2 & \text{if} \quad x = 0 \\ 1/2 & \text{if} \quad x = 1 \\ 0 & \text{otherwise.} \end{cases}$$

Cumulative Distribution Functions (for Cont. RVs)

• For a real-valued random variable (i.e., when $X(\omega)$ is a number), a **cumulative** distribution function (CDF) is a function

$$F_X: \mathbb{R} \mapsto [0,1]$$
 with $F_X(x) = \mathbb{P}(X \le x)$.

- The CDF allows us to compute the probability of any event in \mathcal{F} .
- For example, if we know the CDF, we can calculate

$$\mathbb{P}(1 \le X \le 2.7) = F_X(2.7) - F_X(1)$$

Properties of CDFs

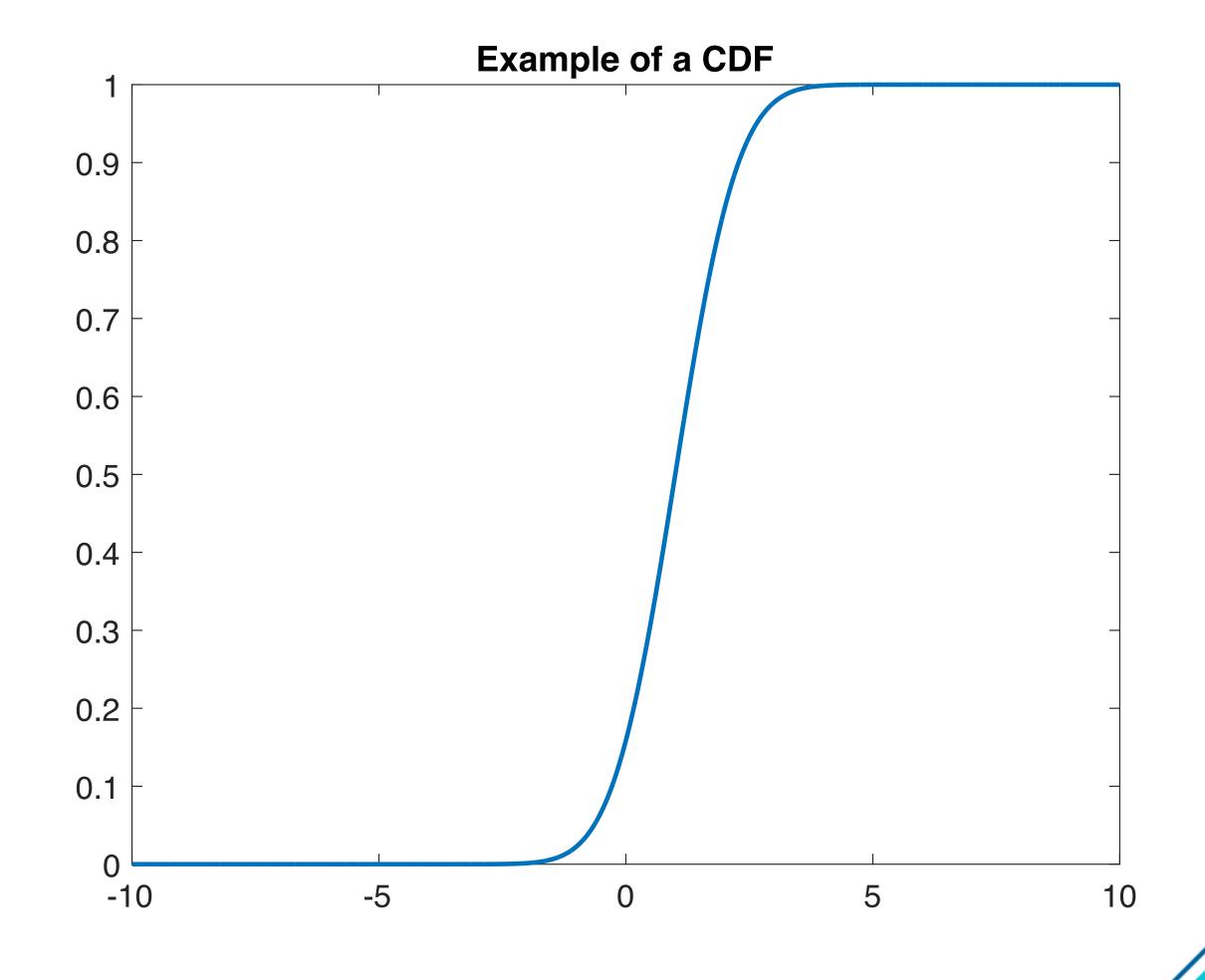
• CDFs satisfy important properties (as can be deduced from the definition and the axioms of probability):

$$-0 \le F_X(x) \le 1.$$

$$\lim_{x \to -\infty} F_X(x) = 0.$$

$$\lim_{x \to +\infty} F_X(x) = 1.$$

$$-x \le y \implies F_X(x) \le F_X(y)$$
.



Probability Density Functions (for Cont. RVs)

 For a continuous random variable, with a differentiable CDF, the probability density function (PDF) is the function

$$f_X(x) = \frac{d}{dx} F_X(x) .$$

- So, it is the derivative of the CDF.
- Like the CDF, it can help us calculate probabilities:

$$\mathbb{P}(a \le X \le b) = \int_{a}^{b} f_X(x) dx$$

The PDF and CDF also satisfy

$$F_X(x) = \int_{-\infty}^{x} f_X(z) dz$$

Properties of PDFs

- PDFs satisfy important properties:
 - $f_X(x) \ge 0.$

$$-\int_{-\infty}^{\infty} f_X(x)dx = 1$$

- For a set
$$A$$
,
$$\int_A f_X(x) dx = \mathbb{P}(X \in A)$$

- Important: In general $f_X(x) \neq \mathbb{P}(X = x)$.
- Exercise: Construct an example for which $f_X(x) \neq \mathbb{P}(X = x)$

