Statistics for Data Science - 2

Week 4 Notes

Continuous Random Variables

1. Cumulative distribution function:

A function $F: \mathbb{R} \to [0,1]$ is said to be a Cumulative Distribution Function (CDF) if

- (i) F is a non-decreasing function taking values between 0 and 1.
- (ii) As $x \to -\infty$, $F \to 0$
- (iii) As $x \to \infty$, $F \to 1$
- (iv) Technical: F is continuous from the right.

2. CDF of a random variable:

Cumulative distribution function of a random variable X is a function $F_X : R \to [0, 1]$ defined as

$$F_X(x) = P(X \le x)$$

Properties of CDF

- $F_X(b) F_X(a) = P(a < X \le b)$
- F_X is a non-decreasing function of x.
- F_X takes non-negative values.
- As $x \to -\infty$, $F_X(x) \to 0$
- As $x \to \infty$, $F_X(x) \to 1$

3. Theorem: Random variable with CDF F(x)

Given a valid CDF F(x), there exists a random variable X taking values in \mathbb{R} such that

$$P(X \le x) = F(x)$$

• If F is not continuous at x and F(X) rises from F_1 to F_2 at x (jump at x), then

$$P(X=x) = F_2 - F_1$$

• If F is continuous at x, then

$$P(X=x)=0$$

4. Continuous random variable:

A random variable X with CDF $F_X(x)$ is said to be a continuous random variable if $F_X(x)$ is continuous at every x.

Properties of continuous random variables

- CDF has no jumps or steps.
- P(X = x) = 0 for all x.

• Probability of X falling in an interval will be nonzero

$$P(a < X \le b) = F(b) - F(a)$$

• Since P(X = a) = 0 and P(X = b) = 0, we have

$$P(a \le X \le b) = P(a < X \le b) = P(a \le X < b) = P(a < X < b)$$

5. Probability density function (PDF):

A continuous random variable X with CDF $F_X(x)$ is said to have a PDF $f_X(x)$ if, for all x_0 ,

$$F_X(x_0) = \int_{-\infty}^{x_0} f_X(x) dx$$

- CDF is the integral of the PDF.
- Derivative of the CDF (wherever it exists) is usually taken as the PDF.
- Value of PDF around $f_X(x_0)$ is related to X taking a value around x_0 .
- Higher the PDF, higher the chance that X lies there.
- 6. For a random variable X with PDF f_X , an event A is a subset of the real line and its probability is computed as

$$P(A) = \int_{A} f_X(x) dx$$

•
$$P(a < X < b) = F_X(b) - F_X(a) = \int_a^b f_X(x) dx$$

7. Density function:

A function $f: \mathbb{R} \to \mathbb{R}$ is said to be a density function if

- (i) $f(x) \ge 0$ (ii) $\int_{-\infty}^{\infty} f_X(x) dx = 1$
- (iii) f(x) is piece-wise continuous
- 8. Given a density function f, there is a continuous random variable X with PDF as f.

9. Support of random variable X

Support of the random variable X with PDF f_X is

$$supp(X) = \{x : f_X(x) > 0\}$$

• supp(X) contains intervals in which X can fall with positive probability.

10. Continuous Uniform distribution:

- $X \sim \text{Uniform}[a, b]$
- PDF:

$$f_X(x) = \begin{cases} \frac{1}{b-a} & a < x < b \\ 0 & \text{otherwise} \end{cases}$$

• CDF:

$$F_X(x) = \begin{cases} 0 & x \le a \\ \frac{x-a}{b-a} & a < x < b \\ 1 & x \ge b \end{cases}$$

11. Exponential distribution:

- $X \sim \text{Exp}(\lambda)$
- PDF:

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

• CDF:

$$F_X(x) = \begin{cases} 0 & x \le 0\\ 1 - e^{-\lambda x} & x > 0 \end{cases}$$

12. Normal distribution:

- $X \sim \text{Normal}[\mu, \sigma^2]$
- PDF:

$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(\frac{-(x-\mu)^2}{2\sigma^2}\right) \qquad -\infty < x < \infty$$

• CDF:

$$F_X(x) = \int_{-\infty}^x f_X(u) du$$

- CDF has no closed form expression.
- Standard normal: Z = Normal(0, 1)

- PDF:
$$f_Z(z) = \frac{1}{\sqrt{2\pi}} \exp\left(\frac{-z^2}{2}\right)$$
 $-\infty < z < \infty$

13. Standardization:

If $X \sim \text{Normal}(\mu, \sigma^2)$, then

$$\frac{X - \mu}{\sigma} = Z \sim \text{Normal}(0, 1)$$

14. To compute the probabilities of the normal distribution, convert probability computation to that of a standard normal.