

## Lecture #3: Expectation

\* The expected value of a random variable is simply the weighted average by the probability!

$$\Rightarrow \mathbb{E}[X] = \int_{-\infty}^{\infty} x f(x) dx \quad \left| \quad \begin{array}{l} \mathbb{E}[X] = \sum_{\omega \in \Omega} X(\omega) P(\{\omega\}) \\ \\ = \sum_{x} x P(X=x). \end{array} \right.$$

Continuous r.v.                                    discrete r.v.

### Example:

Let  $X$  be a random variable with an exponential distribution with parameter  $\lambda > 0$ . Compute  $\mathbb{E}[X]$ .

$$\begin{aligned} \mathbb{E}[X] &= \int_0^{\infty} x \lambda e^{-\lambda x} dx \\ &= -x e^{-\lambda x} \Big|_0^{\infty} + \int_0^{\infty} e^{-\lambda x} dx \\ &= 1/\lambda. \end{aligned}$$

### Important Expectations

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

$$= \int_{-\infty}^{\infty} (x - \mathbb{E}[X])^2 f(x) dx$$

$$= \int_{-\infty}^{\infty} (x^2 - 2x \mathbb{E}[X] + \mathbb{E}[X]^2) f(x) dx$$

$$= \int_{-\infty}^{\infty} x^2 f(x) dx - 2 \mathbb{E}[X] \int_{-\infty}^{\infty} x f(x) dx + \mathbb{E}[X]^2 \int_{-\infty}^{\infty} f(x) dx$$

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

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

2. Moments: The  $n$ -th moment is  $\mathbb{E}[X^n] = \int_{-\infty}^{\infty} x^n f(x) dx$ .

3. Probability:  $\mathbb{E}[1_B] = \int_{-\infty}^{\infty} 1_B(x) f(x) dx$

$$= \int_B f(x) dx$$

$$= P(X \in B).$$

4. Moment Generating Function: Let  $\lambda \in \mathbb{R}$ . The moment generating function is

$$\begin{aligned}\phi(\lambda) &= \mathbb{E}[e^{\lambda X}] \\ &= \int_{-\infty}^{\infty} e^{\lambda x} f(x) dx \\ &= \int_{-\infty}^{\infty} \left(1 + \lambda x + \frac{\lambda^2}{2} x^2 + \frac{\lambda^3}{3!} x^3 + \dots\right) f(x) dx \\ \Rightarrow \phi^{(n)}(0) &= \mathbb{E}[X^n].\end{aligned}$$

## Inequalities

1. Markov's Inequality: Let  $X$  be a positive random variable on  $(\Omega, \mathcal{F}, P)$ . Then for any  $a > 0$  we have

$$P(X > a) \leq \frac{1}{a} \mathbb{E}[X]$$

Proof:

$$\begin{aligned}\mathbb{E}[X] &= \int_0^{\infty} x f(x) dx \\ &= \int_0^a x f(x) dx + \int_a^{\infty} x f(x) dx \\ &\geq \int_a^{\infty} x f(x) dx \\ &\geq a \int_a^{\infty} f(x) dx \\ &= a P(X \geq a).\end{aligned}$$

2. Chebyshev's Inequality: For any random variable on  $(\Omega, \mathcal{F}, P)$

$$P(|X| > a) \leq \frac{1}{a^2} \mathbb{E}[X^2]$$

Proof:

$$P(|X| > a) = P(X^2 > a^2) \leq \frac{1}{a^2} \mathbb{E}[X^2]$$

$$\therefore \Rightarrow P(|X - \mathbb{E}[X]| > n\sigma) \leq \frac{1}{n^2 \sigma^2} \mathbb{E}[(X - \mathbb{E}[X])^2] = \frac{1}{n^2}.$$

3. Chernoff bound:  $P(X > a) \leq e^{-\lambda a} \mathbb{E}[e^{\lambda X}]$

Proof:

$$P(X > a) = P(e^{\lambda X} > e^{\lambda a}) \leq \frac{1}{e^{\lambda a}} \mathbb{E}[e^{\lambda X}].$$