

Chapter 2

- **Permutations:** ${}_nP_r = \frac{n!}{(n-r)!}$, $\binom{n}{r} = \frac{n!}{r!(n-r)!}$
 - **Additive rule:** $P(A \cup B) = P(A) + P(B) - P(A \cap B)$
 - **Conditional probability:** $P(B|A) = \frac{P(A \cap B)}{P(A)}$,
 - **Independence:** $P(A|B) = P(A) \implies P(A \cap B) = P(A)P(B)$
 - **Total probability:** $P(A) = \sum_{n=1}^k P(B_n \cap A) = \sum_{n=1}^k P(B_n)P(A|B_n)$
 - **Bayes' Rule:** \mathcal{P} of even in a partitioned Ω , $P(B_r|A) = \frac{P(B_r)P(A|B_r)}{\sum_{n=1}^{\infty} P(B_n)P(A|B_n)}$
-

Chapter 3

- **Probability mass function:** describes the probability that a **discrete** random variable is exactly equal to some value, i.e.,

$$p : R \rightarrow [0, 1] \quad p(x) = \mathcal{P}(X = x) \iff p(x) \geq 0, \sum_i p(x_i) = 1$$

- **Probability density function:** describes relative probabilities for a set of exclusive **continuous** events, i.e.,

$$\mathcal{P}(a \leq X \leq b) = \int_a^b f(x) dx \iff f(x) \geq 0, \forall x \in \mathbb{R}, \int_{-\infty}^{\infty} f(x) dx = 1$$

- **Cumulative density function:** the sum of continuous probabilities up to a particular point (CDF can be > 1), i.e.,

$$f(x) = \int_{-\infty}^x f(u) du \implies \sum_{i=1}^a p(x_i)$$

Chapter 4

- **Expected value** $E[X]$, \bar{X} : a generalized weighted average, essentially the arithmetic mean of a large number of realizations of some random variable X .

$$\text{Discrete: } E[X] = \sum_x x f(x), \quad \text{Continuous: } E[X] = \int_{-\infty}^{\infty} x f(x) dx$$

- Note: x can be a probability function.
- Applying the mean value μ allows for expected variance, i.e., $\sigma^2 = E[(X - \mu)^2]$