

Discrete Probability

CMPS/MATH 2170: Discrete Mathematics

Applications of Probability in Computer Science

- Algorithms
- Complexity
- Machine learning
- Combinatorics
- Networking
- Cryptography
- Information theory
- ...

Agenda

- Discrete Probability Law (7.1,7.2)
- Independence (7.2)
- Random Variables (7.2)
- Expected Value (7.4)

Examples



- Ex. 1: consider rolling a pair of 6-sided **fair** dice
 - Sample space $\Omega = \{(i, j): i, j = 1, 2, 3, 4, 5, 6\}$
 - Each outcome has the same probability of $1/36$
 - What is the probability that the sum of the rolls is 6?

Let A denote the **event** that the sum of the rolls is 6

$$A = \{(1,5), (5,1), (2,4), (4,2), (3,3)\}$$

$$P(A) = 5/36$$

Examples



- Ex. 2: consider rolling a 6-sided **biased (loaded)** die

- Sample space $\Omega = \{1, 2, 3, 4, 5, 6\}$

- Assume $P(3) = \frac{2}{7}, P(1) = P(2) = P(4) = P(5) = P(6) = \frac{1}{7}$

- What is the probability of getting an odd number?

Let B denote the event of getting an odd number

$$B = \{1, 3, 5\}$$

$$P(B) = \frac{1}{7} + \frac{2}{7} + \frac{1}{7} = \frac{4}{7}$$

Experiment and Sample Space

- **Experiment**: a procedure that yields one of a given set of possible outcomes
 - Ex: flip a coin, roll two dice, draw five cards from a deck, etc.
- **Sample space Ω** : the set of possible outcomes
 - We focus on **countable** sample space: Ω is finite or countably infinite
 - In many applications, Ω is uncountable (e.g., a subset of \mathbb{R})
- **Event**: a subset of the sample space
 - Probability is assigned to events
 - For an event $A \subseteq \Omega$, its probability is denoted by **$P(A)$**
 - Describes beliefs about likelihood of outcomes

Discrete Probability

- Discrete Probability Law

- A function $P: \mathcal{P}(\Omega) \rightarrow [0,1]$ that assigns probability to events such that:

- $0 \leq P(\{s\}) \leq 1$ for all $s \in \Omega$ (Nonnegativity)

- $P(A) = \sum_{s \in A} P(\{s\})$ for all $A \subseteq \Omega$ (Additivity)

- $P(\Omega) = \sum_{s \in \Omega} P(\{s\}) = 1$ (Normalization)

- Discrete uniform probability law: $|\Omega| = n, P(A) = \frac{|A|}{n} \forall A \subseteq \Omega$

Properties of Probability Laws

- Consider a probability law, and let A , B , and C be events
 - If $A \subseteq B$, then $P(A) \leq P(B)$
 - $P(\overline{A}) = 1 - P(A)$
 - $P(A \cup B) = P(A) + P(B) - P(A \cap B)$
 - $P(A \cup B) = P(A) + P(B)$ if A and B are disjoint, i.e., $A \cap B = \emptyset$
- Ex. 3: What is the probability that a positive integer **selected at random** from the set of positive integers not exceeding 100 is divisible by either 2 or 5?

$$\frac{50}{100} + \frac{20}{100} - \frac{10}{100} = 0.6$$

Agenda

- Discrete Probability Law (7.1,7.2)
- **Independence** (7.2)
- Random Variables (7.2)
- Expected Value (7.4)

Independence

- Two events A and B are independent if and only if $P(A \cap B) = P(A) P(B)$
- Ex. 4: Consider an experiment involving two successive rolls of a 4-sided die in which all 16 possible outcomes are equally likely and have probability $1/16$. Are the following pair of events independent?
 - (a) $A = \{ \text{1st roll is 1} \}$, $B = \{ \text{sum of two rolls is 5} \}$ **Yes**
 - (b) $A = \{ \text{1st roll is 4} \}$, $B = \{ \text{sum of two rolls is 4} \}$ **No**

Bernoulli Trials

- **Bernoulli Trial**: an experiment with two possible outcomes
 - E.g., flip a coin results in two possible outcomes: head (H) and tail (T)
- **Independent Bernoulli Trials**: a sequence of Bernoulli trials that are **mutually independent**

Bernoulli Trials

- Ex.5: Consider an experiment involving five **independent** tosses of a biased coin, in which the probability of heads is p .
 - What is the probability of the sequence *HHHTT*?
 - $A_i = \{i\text{-th toss is a head}\}$
 - $P(A_1 \cap A_2 \cap A_3 \cap \bar{A}_4 \cap \bar{A}_5) = P(A_1)P(A_2)P(A_3)P(\bar{A}_4)P(\bar{A}_5) = p^3(1 - p)^2$
 - What is the probability that exactly three heads come up?
 - $P(\text{exactly three heads come up}) = \binom{5}{3}p^3(1 - p)^2$

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Random Variables

- A **random variable (r.v.)** is a **real-valued** function of the experimental outcome.
- Ex. 6: Consider an experiment involving three **independent** tosses of a fair coin.
 - $\Omega = \{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT\}$
 - $X(s)$ = the number of heads that appear for outcome $s \in \Omega$. Then
$$X(HHH) = 3, X(HHT) = X(HTH) = X(THH) = 2,$$
$$X(HTT) = X(THT) = X(TTH) = 1, X(TTT) = 0$$
 - $P(X = 2) = P(\{s \in \Omega: X(s) = 2\}) = P(\{HHT, HTH, THH\}) = 3/8$
 - $P(X < 2) = P(\{HTT, THT, TTH, TTT\}) = 4/8 = 1/2$

Random Variables

- A random variable is a **real-valued** function of the outcome of the experiment.
 - A random variable is called **discrete** if the sample space Ω is **finite or countably infinite**
- We can associate with each random variable certain “averages” of interest, such as the expected value and the variance.

Expected Value

- The **expected value** (also called the **expectation** or the **mean**) of a random variable X on the sample space Ω is equal to

$$E(X) = \sum_{s \in \Omega} X(s) P(\{s\})$$

- Ex. 7: Consider an experiment of tossing a biased coin once where the probability of heads is p .

- $\Omega = \{H, T\}$

- Let X be a r.v. where $X = 1$ if “Head” and $X = 0$ if “Tail”

- $E(X) = 1 \cdot p + 0 \cdot (1 - p) = p$

Expected Value

- Ex. 8: Consider an experiment involving **three independent** tosses of a biased coin in which the probability of heads is p

$$- \Omega = \{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT\}$$

$$- X(s) = \text{the number of heads that appear for outcome } s \in \Omega$$

$$\begin{aligned} - E(X) &= 3 \cdot p^3 + 2 \cdot p^2(1-p) \cdot 3 + 1 \cdot p(1-p)^2 \cdot 3 + 0 \cdot (1-p)^3 \\ &= 3p^3 + 6p^2(1-p) + 3p(1-p)^2 \\ &= 3p \end{aligned}$$

Expected Value

- Ex. 9: Consider an experiment involving n independent tosses of a biased coin in which the probability of heads is p

– $X(s)$ = the number of heads that appear for outcome $s \in \Omega$

$$- E(X) = \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k}$$

$$= np$$