

Information Theory

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Evaluation

- Attendance and class activities 20%
- Midterm Exam 35%
- Final Exam 45%

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Check every week for update

Goals

- Understand the concepts of information entropy and channel capacity
- Understand the digital communication model and its components
- Understand how the components operate
- Understand data compression
- Understand error detection and correction

Course Outline

- Introduction to set theory & probability
- Introduction to information theory
- Coding techniques & data compression
- Information Entropy
- Communication Channel
- Error Detection and Correction

Today's Outline

Introduction to Set Theory and Probability

- 1. Sets, Operations on sets
- 2. Trial, Probability space, Events
- 3. Random variables, Probability distribution
- 4. Expected values, Variance
- 5. Conditional Probability
- 6. Bayes Theory

Sets

Sets

A set is a collection of objects without repetition.
The order of elements is irrelevant.

A set can be expressed by writing all elements

Example : even = { 0, 2, 4, 6, 8, 10 }

OR can be expressed by using a common property of its elements

Example : even = { x : 0 ≤ x ≤ 10 and x is even }

Sets

Elements of sets :

We use the symbols \in and \notin to show that an element belongs to a set or not

Example : even = { 0, 2, 4, 6, 8, 10 }

2 \in even

3 \notin even

Subsets : A subset of a set A is a collection of elements that all belongs to A

Example : S = {2,4} \Rightarrow S \subset even

T = {1, 4} \Rightarrow T $\not\subset$ even

Empty Set : is the set of no elements $\phi = \{ \}$

Operations on sets

Set Union

$$A \cup B = \{ x \mid x \in A \text{ or } x \in B \}$$

Commutative $A \cup B = B \cup A$

Example : even = {0, 2, 4, 6}

odd = {1, 3, 5}

$$\text{even} \cup \text{odd} = \{0, 1, 2, 3, 4, 5, 6\}$$

Operations on sets

Set Intersection

$$A \cap B = \{ x \mid x \in A \text{ and } x \in B \}$$

Commutative $A \cap B = B \cap A$

Example: even = {0, 2, 4, 6}

odd = {1, 3, 5}

$$\text{even} \cap \text{odd} = \phi$$

$$\text{even} \cap \{1, 2, 3\} = \{2\}$$

Operations on sets

Set Subtraction

$$A - B = \{ x \mid x \in A \text{ and } x \notin B \}$$

Non commutative $A - B \neq B - A$

Example : even = {0, 2, 4, 6}

odd = {1, 3, 5}

$$\text{even} - \text{odd} = \text{even}$$

$$\text{even} - \{1, 2, 3\} = \{0, 4, 6\}$$

Operations on sets

Power Set

Power Set: is the set of all subsets

$$P(A) = 2^A = \{ B \mid B \subseteq A \}$$

Example: odd = {1, 3, 5}

$$P(\text{odd}) = 2^{\text{odd}} = \{ \phi, \{1\}, \{3\}, \{5\}, \{1,3\}, \{1,5\}, \{3,5\}, \text{odd} \}$$

Cardinality

Cardinality is the number (#) of elements in a set

Example: # (odd) = 3 , # (2^{odd}) = 8 = 2³ = 2^{#(odd)}

Trials

Many problems in probabilities and statistics involve situations in which an experiment with the possible outcomes is repeated several times.

Each repetition of the experiment is called a *trial*.

Example : throw a dice



Probability

Probability:

A probability is a number associated with or assigned to a set in order to measure it in some sense

Probability Space:

Consider the symbols Ω and P where

Ω : universal set ;

P : is a probability function mapping power set to reals in the interval $[0,1]$; i.e. $P: 2^\Omega \rightarrow [0,1]$

With the following 3 properties:

- (1) $P(\emptyset) = 0$
- (2) $P(\Omega) = 1$
- (3) $\forall A = \{a_1, a_2, \dots, a_m\} \subset \Omega,$

$$P(A) = \sum_{n=1}^m P(a_n),$$

where $P(a_n) = P(\{a_n\})$

Event

The subset of Ω to which a probability has been assigned is called an event

An event is a classification of trial outcome.

Example :



The set of events of throwing a dice is $\{1,2,3,4,5,6\}$

Random Variables

Random Variable is a numerical quantity whose value depends on chance.

Example : A person to be selected at random

For a universal set Ω and a $a \in \Omega$

The function $X: \Omega \rightarrow$ set of numerical values is called a random variable

Example : Throwing 2 dice we get



$$\Omega = \{ (m,n) \mid m,n \in \{1,2,\dots,6\} \}$$

(1,1), (1,2), (1,3), (1,4), (1,5), (1,6)
 (2,1), (2,2), (2,3), (2,4), (2,5), (2,6)
 $\Omega = \{ (3,1), (3,2), (3,3), (3,4), (3,5), (3,6)$
 (4,1), (4,2), (4,3), (4,4), (4,5), (4,6)
 (5,1), (5,2), (5,3), (5,4), (5,5), (5,6)
 (6,1), (6,2), (6,3), (6,4), (6,5), (6,6)

Random Variables

Proposition

If X and Y are random variables then $X+Y$, $X-Y$, XY and X/Y ($Y \neq 0$) are random variables

Probability Distribution

The probability that the random variable X takes the value a_n is

$$p_n = P(X = a_n)$$

$$p_n \geq 0, \sum p_n = 1$$

If certainty is 100% then probability = 1 (unity)

In general $\forall n \quad 0 \leq p_n \leq 1$

Example : fair coin front \uparrow



back \downarrow



$\left(\begin{array}{c} \uparrow \downarrow \\ \frac{1}{2} \quad \frac{1}{2} \end{array} \right)$

Probability Distribution

Example : fair dice

$$\left(\begin{array}{cccccc} 1 & 2 & 3 & 4 & 5 & 6 \\ 1/6 & 1/6 & 1/6 & 1/6 & 1/6 & 1/6 \end{array} \right)$$

Example of random variable : sum of two dice throws

Consider the random variable X_1 as outcome of the first dice:
 $X_1((m, n)) = m$

Consider the random variable X_2 as outcome of the second dice:
 $X_2((m, n)) = n$

Consider the random variable X_3 as sum of X_1 and X_2 ($X_3 = X_1 + X_2$):
 $X_3((m, n)) = m + n$

	+	1	2	3	4	5	6	
1		2	3	4	5	6	7	Probability distribution of X_1
2		3	4	5	6	7	8	
3		4	5	6	7	8	9	Probability distribution of X_2
4		5	6	7	8	9	10	
5		6	7	8	9	10	11	Probability distribution of X_3
6		7	8	9	10	11	12	

Expected value (mean or average value)

The Expected value is designed as follows :

Take the value of (the random variable) X at each point $a \in \Omega$, multiply it by the probability of that point (i.e $P(a)$) and sum over all $a \in \Omega$

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

Note that:

$$E(X + Y) = E(X) + E(Y)$$

$$E(cX) = c E(X) \quad (\text{for constant } c)$$

Expected value (mean or average value)

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

Example

Consider a land with 100 acre with 50% of price \$150 per acre, 30% of price \$100 per acre and 20% of price \$50 per acre.

What is the average (expected or mean) price per acre for the whole land?

30%	20%
\$100	\$50
50%	\$150

Answer:

$$\Omega = \{50\%, 30\%, 20\%\} = \{50/100, 30/100, 20/100\}$$

$$E(X) = 150 * 50/100 + 100 * 30/100 + 50 * 20/100 = 115 \$ \text{ per acre}$$

Variance

The Variance of a variable X (denoted by $V(X)$ or $\sigma^2(X)$) is defined by

$$V(X) = E[(X - E(X))^2]$$

Note that: for a constant c

$$\begin{aligned} V(c) &= E[(c - E(c))^2] \\ &= E[(c - c)^2] \\ &= 0 \end{aligned}$$

$$\begin{aligned} V(cX) &= E[(cX - E(cX))^2] \\ &= E[(cX - cE(X))^2] \\ &= E[c^2(X - E(X))^2] \\ &= c^2 E[(X - E(X))^2] \\ &= c^2 V(X) \end{aligned}$$

Variance

Example:

For a fair coin toss (with front denoted by \uparrow and back denoted by \downarrow) we have

	Ω				
	(\uparrow, \uparrow)	(\uparrow, \downarrow)	(\downarrow, \uparrow)	(\downarrow, \downarrow)	Σ
p	1/4	1/4	1/4	1/4	1
$x = \#(\uparrow)$	2	1	1	0	4
$x P(x)$	2/4	1/4	1/4	0/4	$E[X] = 1$
$(x - E[x])^2 P(x)$	(2-1) ² /4	(1-1) ² /4	(1-1) ² /4	(0-1) ² /4	$V[X] = 1/2$

Exercise

Prove this equality

$$V(X) = E(X^2) - E(X)^2$$

Exercise: Answer

Prove this equality $V(X) = E(X^2) - E(X)^2$

Answer:

$$\begin{aligned} V(X) &= E[(X - E(X))^2] \\ &= E[(X - E(X))(X - E(X))] \\ &= E[X^2 - 2XE(X) + E(X)^2] \\ &= E(X^2) - 2E(X)^2 + E(X)^2 \\ &= E(X^2) - E(X)^2 \end{aligned}$$

Conditional Probability

Conditional probability is the probability of A when B is known

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Example:



If a dice is rolled:

Let an event A = the dice comes up with 5 = {5}

Let an event B = the dice comes up odd = {1, 3, 5}

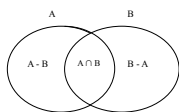
Then: $P(A) = 1/6$, $P(B) = P(\{1, 3, 5\}) = P(\{1\}) + P(\{3\}) + P(\{5\}) = 1/6 + 1/6 + 1/6 = 1/2$

$A \cap B = \{5\}$, $P(A \cap B) = P(\{5\}) = 1/6$

Hence $P(A|B) = P(A \cap B) / P(B) = (1/6) / (1/2) = 1/3$

Conditional Probability

Note that :



$$P(B) = P(A \cap B) + P(B - A) \geq P(A \cap B) \Rightarrow P(A|B) \leq 1$$

$$\text{If } B \subseteq A \Rightarrow A \cap B = B \Rightarrow P(A|B) = 1$$

Bayes Theorem

$$P(B|A) = \frac{P(B)P(A|B)}{P(A)}$$

Knowing the outcome of a particular situation, one (using Bayes theorem) can find the probability that the outcome occurred as a result of a particular previous event.

Note that:

- in general $P(A|B) \neq P(B|A)$
- $P(A|B) = P(B|A)$ only if $P(A) = P(B)$

Bayes Theorem

Example: Throw a single dice: let $A = \{x | x \text{ odd}\}$, $B = \{1, 2\}$



$$P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{1/6}{2/6} = 1/2$$

$$P(B|A) = \frac{P(B)P(A|B)}{P(A)} = \frac{(2/6)(1/2)}{3/6} = 1/3$$

Bayes Theorem

Properties:

$$1. P(\phi|B) = \frac{P(\phi \cap B)}{P(B)} = \frac{P(\phi)}{P(B)} = \frac{0}{P(B)} = 0$$

$$2. P(\Omega|B) = \frac{P(\Omega \cap B)}{P(B)} = \frac{P(B)}{P(B)} = 1$$

$$3. P(A|B) = \frac{P(A \cap B)}{P(B)} = \frac{\sum_{a \in A} P(\{a\} \cap B)}{P(B)} = \frac{P(\{a\} \cap B)}{P(B)} = \sum_{a \in A} P(\{a\} | B)$$

Independence of Random Variables

Two discrete random variables, X_1 and X_2 are independent if

$$\forall \text{ value } j, k \quad P(X_1 = j \cap X_2 = k) = P(X_1 = j) P(X_2 = k)$$

Note that:

If X_1 and X_2 are independent random variables then it is easy to show that:

$$P(X_1 = j | X_2) = P(X_1 = j)$$

END