

# Information Theory & Coding

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# CHAPTER-3

## Probability Theory



## Review of Probability Concept

- Probability means possibility or how likely an event is to occur, or how likely it is that a proposition is true or how likely something is to happen.
- The concept of probability is used to predict the occurrences of yes or no from an event.
- For example the possibility of getting head or tail from the tossing the coin. We say that the probability of the coin landing H is  $\frac{1}{2}$ , and the probability of the coin landing T is  $\frac{1}{2}$ .



## Review of Probability Concept

- When a single die is thrown, there are six possible outcomes: 1, 2, 3, 4, 5, 6. The probability of any one of them is  $1/6$ .
- In general the formula of probability is defined as,

$$\text{Probability of an event (P)} = \frac{\text{Number of favourable outcomes}}{\text{Total number of outcomes}}$$

- There are 5 bottles in a bag: 4 are blue, and 1 is red. What is the probability that a blue bottle gets picked?  $[P=4/5=0.8]$



## Concept of random variable

- “A random variable is a variable that can take on different values randomly”.
- Measuring a possible values from random experiment. For example tossing a coins, predicting a rain for tomorrow.
- Random variable is a different from simple algebra variable, for an example of algebra variable is  $x+2=9$ .



## Concept of random variable

- Put the value of  $x$  as 5 and you will get the result.
- But the random variable is different from Algebra variable. A Random Variable has a whole set of value and it could take on any of those values, randomly.
- $X = \{0, 1, 2\}$ .  $X$  could be 0, 1, or 2 randomly. And they might each have a different probability.



## Concept of random variable

Suppose that a coin is tossed twice so that the sample space is  $S = \{HH, HT, TH, TT\}$ . Let  $X$  represent the number of heads that can come up. The value of  $X$  could be 0, 1, or 2.

**Table 3.1**

Sample	HH	HT	TH	TT
$X$	2	1	1	0



## Concept of random variable

A random variable can be classified into two types, a random variable that takes on a finite or countable infinite number of values is called a discrete random variable while one which takes on a non-countable infinite number of values is called continuous random variables.





## Function of random variable

If  $X$  is a random variable and  $Y=g(X)$ , then  $Y$  itself is a random variable.

### ❖ Distribution and density function Moments

➤ **Discrete Probability Distributions:** Suppose a random variable  $X$  may take  $k$  different values, with the probability that  $X = x_i$  defined to be  $P(X = x_i) = p_i$ .

The probabilities  $p_i$  must satisfy the following:

- $0 < p_i < 1$  for each  $i$
- $p_1 + p_2 + \dots + p_k = 1$ .
- A probability distribution over discrete variables may be described using a probability mass function (PMF).



## Function of random variable

➤ Mean of discrete random variable:

$$\begin{aligned}\mu_X &= x_1 p_1 + x_2 p_2 + \cdots + x_k p_k \\ &= \sum x_i p_i\end{aligned}$$

If  $X$  is a random variable with mean  $E(X) = \mu$ , then the variance of  $X$  is defined by

$$\text{Var}(X) = E((X - \mu)^2).$$

$$\text{Var}(X) = E(X^2) - E(X)^2$$



## Function of random variable

- Suppose that a coin is tossed twice so that the sample space is  $S = \{HH, HT, TH, TT\}$ . Find the probability function corresponding to the random variable  $X$ .

$$P(X = 2) = P(HH) = \frac{1}{4}$$

$$P(X = 1) = P(HT) = \frac{1}{4}$$

$$P(X = 1) = P(TH) = \frac{1}{4}$$

$$P(X = 0) = P(TT) = \frac{1}{4}$$

$$P(X = 1) = P(HT) = \frac{1}{4} + P(X = 1) = P(TH) = \frac{1}{4} = \frac{1}{2}$$



# Function of random variable

## ❖ Continuous Random Variables:

When working with continuous random variables, we describe probability distributions using a probability density function (PDF).

$$\Pr[a \leq X \leq b] = \int_a^b f_X(x) dx.$$

$$\int_{-\infty}^{\infty} f_X(x) dx = 1.$$



## Function of random variable

Let  $X$  be a continuous random variable with range  $[a, b]$  and probability density function  $f(x)$ . The expected value of  $X$  is defined by,

$$E(X) = \int_a^b x f(x) dx.$$

$$E(X) = \sum_{i=1}^n x_i p(x_i).$$



## Function of random variable

### ❖ Uniform Distribution:

X is uniform on  $[a, b]$  if X is equally likely to take any value in the range from a to b. The value of area under the curve is 1.0

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{for } a \leq x \leq b \\ 0 & \text{for all other values} \end{cases}$$



# Characteristic function and conditional statistics

- A characteristic function is simply the Fourier transform, in probabilistic language. The characteristic function always exists when treated as a function of a real-valued.
- Function is called real-valued if it corresponds to a random variable that is symmetric around the origin.

$$\phi_X(t) = E \exp(it^\top X)$$

This function is called characteristic function (cf) of X



# Characteristic function and conditional statistics

- Probability of an event occurring given that another event has already occurred is called a conditional probability.

$$P(B|A) = P(A \text{ and } B) / P(A)$$

The probability that event B occurs, given that event A has already occurred.





## Sequence of Random Variables

- A random sequence or a discrete-time random process is a sequence of random variables  $X_1, \dots, X_n$ . For sample space  $S$ ,

$$S = \{s_1, s_2, \dots, s_k\}.$$

Then, a random variable  $X$  is a mapping that assigns a real number to any of the possible outcomes  $s_i$ ,  $i=1,2,\dots,k$ . Thus, we may write

$$X(s_i) = x_i, \quad \text{for } i = 1, 2, \dots, k.$$



## Rayleigh distribution

- It is a continuous probability distribution for nonnegative-valued random variables. The probability density function and cumulative distribution are represented as:

$$p(x) = \frac{x}{q^2} \exp\left(-\frac{x^2}{2q^2}\right)$$

$$F(x) = 1 - \exp\left(-\frac{x^2}{2q^2}\right)$$



## Rice Distribution

- A Rice distribution is one way to model the paths scattered signals take to a receiver. Specifically, this distribution models line-of-sight scatter (FM, radio waves, microwaves) — transmissions between two stations in view of each other that have an unobstructed path between them.



# Lognormal Distribution

- It is a continuous probability distribution of a random variable whose logarithm is normally distributed. If the random variable  $X$  is log-normally distributed, then  $Y = \ln(X)$  has a normal distribution.
- Amounts of rainfall, Size distributions of rainfall droplets, the volume of gas in a petroleum reserve are the some natural phenomenon where the used of lognormal distribution can be seen..



# Poisson distribution

- It is a discrete probability distribution that expresses the probability of a given number of events occurring in a fixed interval of time.
- For an example is the number of diners in a certain restaurant every day. If the average number of diners for seven days is 500, you can predict the probability of a certain day having more customers.
- The number of cases of a disease in different towns.
- The Poisson distribution can be applied to systems with a large number of possible events, each of which is rare.





## Poisson distribution

$$P(X = x) = e^{-\lambda} \frac{\lambda^x}{x!} \quad x = 0, 1, 2, 3, 4, \dots$$

Where X=Random variable

$\lambda$ = mean or rare event

e=exponential constant=2.7





## Poisson distribution

Using the Poisson distribution  $P(X = r) = e^{-\lambda} \frac{\lambda^r}{r!}$  write down the formulae for  $P(X = 0)$ ,  $P(X = 1)$ ,  $P(X = 2)$  and  $P(X = 6)$ , noting that  $0! = 1$ .

$$P(X = 0) = e^{-\lambda} \times \frac{\lambda^0}{0!} = e^{-\lambda} \times \frac{1}{1} \equiv e^{-\lambda}$$

$$P(X = 2) = e^{-\lambda} \times \frac{\lambda^2}{2!} = \frac{\lambda^2}{2} e^{-\lambda}$$

$$P(X = 1) = e^{-\lambda} \times \frac{\lambda}{1!} = \lambda e^{-\lambda}$$

$$P(X = 6) = e^{-\lambda} \times \frac{\lambda^6}{6!} = \frac{\lambda^6}{720} e^{-\lambda}$$





## Poisson distribution

Mean (Expectation) and Variance of Poisson Distribution is given that:

$$\mu = \lambda$$

$$\sigma = \sqrt{\lambda}$$

$$E(X) = np \quad \text{and} \quad V(X) = npq \approx np$$







## References

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