Basics of Signals and Systems

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Random Variables and Random Processes

Some basics of probability theory are discussed before going to random variables.

Basics of Probability Theory

Probability of an event A represented by P(A) is given by

$$P(A) = \frac{N_A}{N_S}$$

where, N_S is the number of times the experiment is carried and N_A is number of times the event A occured. Probability of any event can be exactly calculated only when the number of experiments is huge ideally infinity. Hence, we generally go for relative probability which is given above. Clearly, $0 \le P(A) \le 1$.

Let 'S' be the sample space having N events $A_1, A_2, A_3, \dots A_N$. Two events are said to be mutually exclusive or statistically independent if $A_i \cap A_j = \phi$ and $\bigcup_{i=1}^N A_i = S$ for all i and j.

Joint Probability:

Joint probability of two events A and B which is defined as the probability of the occurrence of both the events A and B is given by

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

Conditional Probability

Conditional probability of two events A and B represented as P(A/B) and defined as the probability of the occurrence of event A after the occurrence of B is given by

$$P(A/B) = P(A \cap B)/P(B)$$

$$P(B/A) = P(A \cap B)/P(A)$$

$$\implies P(A/B).P(B) = P(B/A).P(A) = P(A \cap B)$$

Chain rule

Let us consider a chain of events $A_1, A_2, A_3, \dots A_N$ which are dependent on each other. Then the probability of occurrence of the sequence

$$P(A_N, A_{N-1}, A_{N-2}, \cdots A_1) = P(A_N/A_{N-1}, A_{N-2}, \cdots A_1) \cdot P(A_{N-1}/A_{N-2}, A_{N-3}, \cdots A_1) \cdot \cdots P(A_2/A_1) \cdot P(A_1/A_{N-2}, A_1/A_{N-2}, \cdots A_1) \cdot \cdots P(A_2/A_1) \cdot P(A_1/A_{N-1}, A_1/A_{N-2}, \cdots A_1) \cdot \cdots P(A_1/A_{N-1}, A_1/A_{N-1}, \cdots A_1/A_{N-1}, \cdots A_1) \cdot \cdots P(A_1/A_{N-1}, A_1/A_{N-1}, \cdots A_1/A_{$$

Bayes Rule

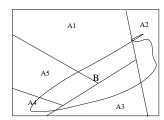


Figure 1: Partition space

In the above figure, if $A_1, A_2, A_3, \dots A_5$ partition the sample space S, then $A_1 \cap B, A_2 \cap B, A_3 \cap B, A_4 \cap B, and A_5 \cap B$ partition B. therefore,

$$P(B) = \sum_{i=1}^{n} P(A_i \cap B)$$
$$= \sum_{i=1}^{n} P(B/A_i).P(A_i)$$

$$P(A_i/B) = P(A_i \cap B)/P(B)$$

$$= \frac{P(B/A_i).P(A_i)}{\sum_{i=1}^{n} P(B/A_i).P(A_i)}$$

In the above equation, $P(A_i/B)$ is called posterior probability, $P(B/A_i)$ is called likelihood, $P(A_i)$ is called prior probability and $\sum_{i=1}^{n} P(B/A_i).P(A_i)$ is called evidence.

Random Variable

Random variable is a function whose domain is sample space and whose range is the set of real numbers.

Probabilistic description of a Random Variable

Cumulative Probability Distribution: It is represented as $F_X(x)$ and defined as

$$F_X(x) = P(X \le x)$$

If $x_1 < x_2$ then $F_X(x_1) < F_X(x_2)$ and $0 \le F_X(x) \le 1$.

Probability Density Function: It is represented as $f_X(x)$ and defined as

$$f_X(x) = \frac{d}{dx} F_X(x)$$

$$\implies P(x_1 \le X \le x_2) = \int_{x_1}^{x_2} f_X(x) dx$$