

# EE306: Signals and Systems II

## Lecture 1

- \* About the course (3 modules).
- \* Applications (wireless communications, data storage, machine learning - internet of things).
- \* Importance of probability theory  
→ What we are going to review.

### Random experiment

It is an experiment in which the outcome varies in a non-deterministic manner when the experiment is repeated under the same conditions.

Ex 1 Roll a dice and note the number on the upper face of it ( $E_1$ ).

Ex 2 Toss a coin 2 times and note the sequence ( $E_2$ ).

Ex 3 Toss a coin 4 times and note # of tails ( $E_3$ ).

Ex 4 Pick a real number at random bet. 0 and 1 ( $E_4$ ).

## Sample space

- \* Sample point: A result that cannot be decomposed into other results (outcome).
- \* Sample space  $S$ : The set of all possible outcomes.

Ex 1  $S(E_1) = \{1, 2, 3, 4, 5, 6\}.$

Ex 2  $S(E_2) = \{HH, HT, TH, TT\}.$

Ex 3  $S(E_3) = \{0, 1, 2, 3, 4\}.$

Ex 4  $S(E_4) = \{x \mid 0 \leq x \leq 1\} = [0, 1]$  (infinite).

Discrete  $S$   
Continuous  $S$

## Events

A situation specified by certain conditions, which occurs when the experiment outcome satisfies these conditions. Thus,

→ An event  $A$  corresponds to a subset of  $S$ .

→ The certain event is  $S$ .

The impossible event is  $\phi$ .

Ex 1 An even number on upper face:

$$A = \{2, 4, 6\}.$$

Ex2 At least one tail observed :

$$A = \{HT, TH, TT\}.$$

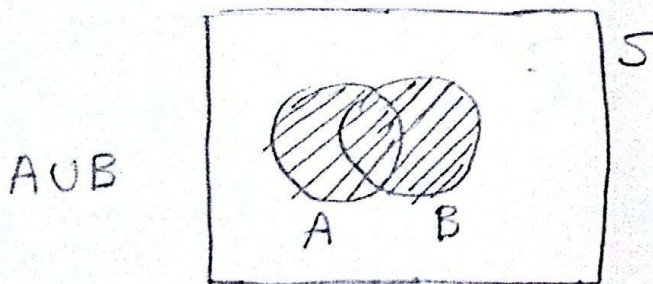
Ex3 Five or more tails observed :  $A = \phi$ .

Ex4 The number is such that  $|x| > 0.9$  :

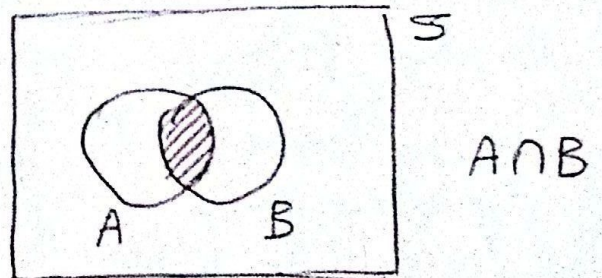
$$A = \{x : 0 \leq x < 0.9\} = [0, 0.9).$$

Basics of set theory Probability : Number assigned to an event, indicating likelihood.

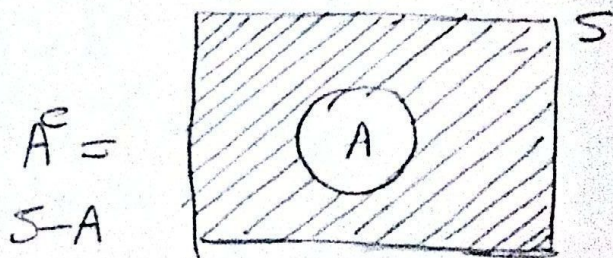
\* Essential notation using Venn diagrams :



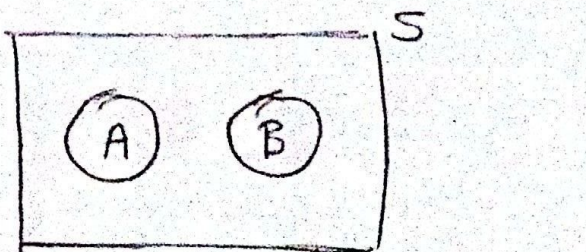
$$\{x : x \in A \text{ or } x \in B\}$$



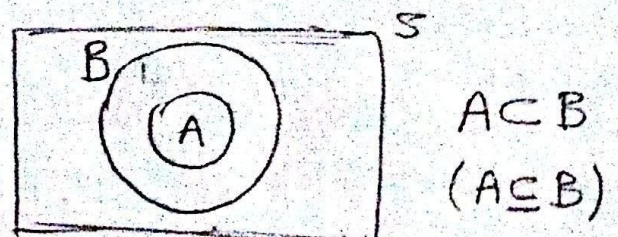
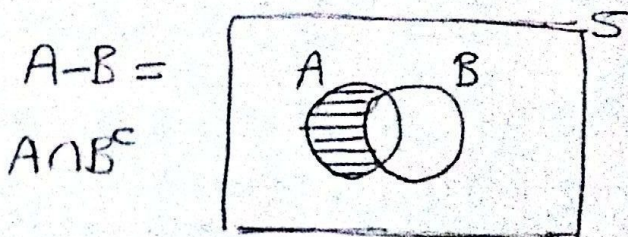
$$\{x : x \in A \text{ and } x \in B\}$$



$$\{x : x \notin A\}$$



$$A \cap B = \phi \text{ (disjoint)}$$





Ex 1 An odd number on upper face 2

$$B = \{1, 3, 5\}. \text{ Thus,}$$

$$A \cup B = \{1, 2, \dots, 6\} = S \text{ and } A \cap B = \phi.$$

Ex 2 Exactly one head observed 2

$$B = \{HT, TH\}. \text{ Thus,}$$

$$A \cup B = \{HT, TH, TT\} = A \text{ because } B \subset A.$$

### Properties of set operations

\* Commutativity 2  $A \cup B = B \cup A, A \cap B = B \cap A.$

\* Associativity 2  $A \cup (B \cap C) = (A \cup B) \cap C.$

$$A \cap (B \cup C) = (A \cap B) \cup C.$$

\* Distributivity 2  $A \cup (B \cap C) = (A \cup B) \cap (A \cup C).$

$$A \cap (B \cup C) = (A \cap B) \cup (A \cap C).$$

Brackets are very important!

\* DeMorgan's rules 2

$$(A \cup B)^c = A^c \cap B^c \text{ and } (A \cap B)^c = A^c \cup B^c.$$

EX 1 Apply DeMorgan to A and B 2

$$A^c = \{1, 3, 5\}, B^c = \{2, 4, 6\} \quad (A \cap B)^c = \phi^c = S = A \cup B^c.$$

\* Set equality:  $A=B \iff ACB$  and  $BCA$ .

\*  $ACB$  means:  $B$  necessary for  $A$   
while  $A$  sufficient for  $B$ .

\*  $\bigcup_{k=1}^n A_k = A_1 \cup A_2 \cup \dots \cup A_n$ ,  $\bigcap_{k=1}^n A_k = A_1 \cap A_2 \cap \dots \cap A_n$ .

### Probability axioms

I.  $P(A) \geq 0$ .      II.  $P(S) = 1$ .

III. If  $A_1, A_2, \dots, A_n$  are pairwise disjoint, i.e.,  
 $A_i \cap A_j = \phi$ ,  $\forall i \neq j$ , then  $P\left(\bigcup_{k=1}^n A_k\right) = \sum_{k=1}^n P(A_k)$ .

Corollary  $P(A^c) = 1 - P(A)$ .

Proof  $A \cap A^c = \phi$ ; thus,  $P(A \cup A^c) = P(A) + P(A^c)$ .

But  $A \cup A^c = S$  and  $P(S) = 1 \implies P(A^c) = 1 - P(A)$ .  $\blacksquare$   
 $\rightarrow \textcircled{1}$

Corollary  $P(A) \leq 1$ .

Proof From  $\textcircled{1}$ ,  $P(A) = 1 - P(A^c)$

since  $P(A^c) \geq 0$ ,  $P(A) \leq 1$ .  $\blacksquare$

Corollary  $P(\phi) = 0$ . Easy to prove.

Corollary  $P(A \cup B) = P(A) + P(B) - P(A \cap B)$ .

Proof We can write  $A \cup B = \overset{E_1}{(A \cap B^c)} \cup \overset{E_2}{(B \cap A^c)} \cup \overset{E_3}{(A \cap B)}$ ,  
where  $E_1, E_2$ , and  $E_3$  are pairwise disjoint.

Thus,  $P(A \cup B) = P(E_1) + P(E_2) + P(E_3)$ .  $\rightarrow$  ②

But  $P(E_1) = P(A) - P(A \cap B)$  and  $P(E_2) = P(B) - P(A \cap B)$ .

Substituting in ② proves the corollary.  $\blacksquare$

Corollary The inclusion-exclusion principle:

$$P\left(\bigcup_{k=1}^n A_k\right) = \sum_{k=1}^n (-1)^{k+1} \sum_{\substack{\{i_1, i_2, \dots, i_k\} \\ \subseteq \{1, 2, \dots, n\}}} P\left(\bigcap_{j=1}^k A_{i_j}\right).$$

Proof By induction using previous corollary (HW).

Corollary If  $A \subseteq B$ , then  $P(A) \leq P(B)$ .

Proof We can write  $B = \overset{E_1}{(B \cap A^c)} \cup \overset{E_2}{(B \cap A)}$ ,  
where  $E_1$  and  $E_2$  are disjoint.

Thus,  $P(B) = P(B \cap A^c) + P(B \cap A)$ .

But  $B \cap A = A \Rightarrow P(A) = P(B) - \underbrace{P(B \cap A^c)}_{\geq 0}$   
 $\Rightarrow P(A) \leq P(B)$ .  $\blacksquare$

# EE 306: Signals and Systems II

## Lecture 2

### Conditional probability

\* For two events A and B:

$$P(A|B) = \frac{P(A \cap B)}{P(B)}, \text{ for } P(B) > 0.$$

The relative frequency of the event  $A \cap B$  in experiments where B occurred, i.e.,  $n_{A \cap B} / n_B$ .

$$* P(A \cap B) = P(A|B) P(B) = P(B|A) P(A).$$

### Ex 1 Binary symmetric channel

characterized as follows:

$A_i$ : Event I/P is  $i$ .

$B_j$ : Event O/P is  $j$ .

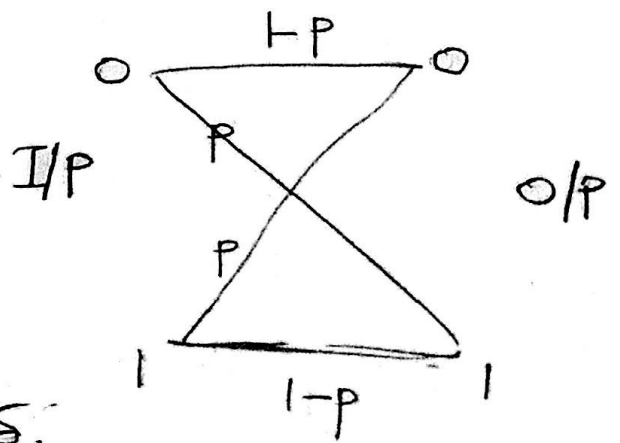
Error occurs with probability

$p$ , and  $P(A_0) = 1 - \delta$ ,  $P(A_1) = \delta$ .

Find  $P(A_i \cap B_j)$ ,  $\forall i, j \in \{0, 1\}$ .

Sol.  $P(A_i \cap B_j) = P(B_j | A_i) P(A_i)$ .

$$P(B_j | A_i) = 1 - p, \quad i = j, \quad \text{and} \quad P(B_j | A_i) = p, \quad i \neq j.$$



Thus,  $P(A_0, B_0) = P(B_0|A_0)P(A_0) = (1-p)(1-s)$ .

Similarly,  $P(A_0, B_1) = p(1-s)$ ,

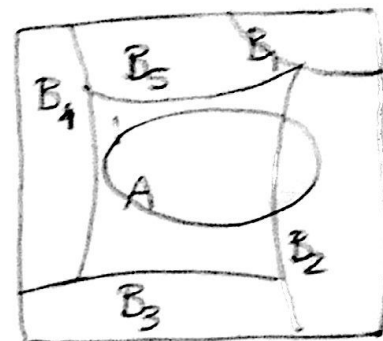
$P(A_1, B_0) = ps$ , and  $P(A_1, B_1) = (1-p)s$ .

### Total probability theorem

\* Suppose  $B_1, B_2, \dots, B_n$  are mutually exclusive events such that  $\bigcup_{k=1}^n B_k = S$ .

$B_k, k \in \{1, 2, \dots, n\}$  are a partition

of  $S$ . Thus,  $P(A) = \sum_{k=1}^n P(A|B_k)P(B_k)$ .



Proof  $A = A \cap S = A \cap \left(\bigcup_{k=1}^n B_k\right)$

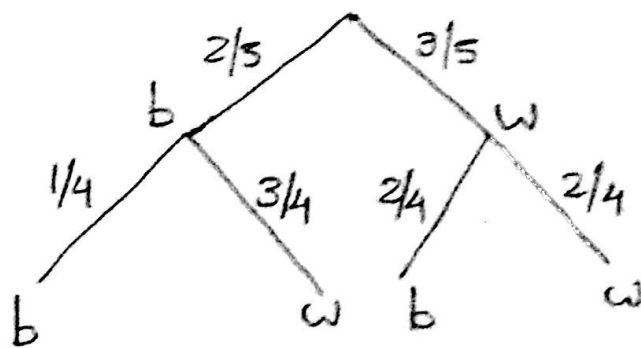
$= (A \cap B_1) \cup (A \cap B_2) \cup \dots \cup (A \cap B_n)$ .

Thus,  $P(A) = \sum_{k=1}^n P(A \cap B_k) = \sum_{k=1}^n P(A|B_k)P(B_k)$ . ▣

Ex 2 An urn with 2 black and 3 white balls. Two balls are selected at random (no replacement) and the color is noted. Find the probability that second ball is black.

Sol. We build this tree:

$$P(B_2) = P(B_2|B_1)P(B_1) + P(B_2|W_1)P(W_1).$$





$$\text{Thus, } P(B_2) = \left(\frac{1}{4}\right)\left(\frac{2}{5}\right) + \left(\frac{2}{4}\right)\left(\frac{3}{5}\right) = \frac{8}{20} = \frac{2}{5}.$$

### Bayes' rule

\* We stick to the same partition of  $S$ . Suppose that event  $A$  occurs. Therefore,

$$P(B_j | A) = \frac{P(B_j \cap A)}{P(A)} = \frac{P(A | B_j) P(B_j)}{\sum_{k=1}^n P(A | B_k) P(B_k)}.$$

→ Very useful when

$P(A | B_k)$  and  $P(B_k)$  are known for all  $k$ .

### Ex 3 Posterior probabilities 2

Consider the binary symmetric comm. channel of Ex 1.

Find  $P(A_0 | B_0)$  and  $P(A_1 | B_0)$ .  $P(\text{tx} | \text{rx})$

$$\begin{aligned} \text{Sol. } P(B_0) &= P(B_0 | A_0) P(A_0) + P(B_0 | A_1) P(A_1) \\ &= (1-p)(1-\delta) + p\delta. \end{aligned}$$

$$P(A_0 | B_0) = \frac{P(B_0 | A_0) P(A_0)}{P(B_0)} = \frac{(1-p)(1-\delta)}{(1-p)(1-\delta) + p\delta}.$$

$$P(A_1 | B_0) = \frac{P(B_0 | A_1) P(A_1)}{P(B_0)} = \frac{p\delta}{(1-p)(1-\delta) + p\delta}.$$

$$\text{If } \delta = \frac{1}{2}, \quad P(A_0 | B_0) = 1-p, \quad P(A_1 | B_0) = p.$$

## Independence of events

\* Events A and B are independent when:

$$P(A|B) = P(A) \text{ or } P(A \cap B) = P(A)P(B).$$

\* Events A, B, and C are independent when:

→ They are pairwise independent.

$$\rightarrow P(A \cap B \cap C) = P(A)P(B)P(C).$$

$$\left[ P(C|A \cap B) = \frac{P(A \cap B \cap C)}{P(A \cap B)} = P(C) \Rightarrow P(A \cap B \cap C) = P(A)P(B)P(C). \right]$$

Notes →  $P(A, B|C) = P(A|B, C)P(B|C).$

$$\rightarrow P(A|C) = \sum_k P(A|B_k, C)P(B_k|C). \quad [ , \Leftrightarrow \cap ]$$

## Counting methods    n objects, k samples

\* Sampling with replacement with ordering:

$$\text{No. of options} = n^k.$$

\* Sampling without replacement with ordering:

$$\text{No. of options} = {}^n P_k = \frac{n!}{(n-k)!}.$$

\* Sampling without replacement without ordering:

$$\text{No. of options} = \binom{n}{k} = {}^n C_k = \frac{n!}{k!(n-k)!}.$$

Idea No. of permutations of  $k$  objects =  $k!$ , and they are all one option.

\* Sampling with replacement without ordering:

$$\text{No. of options} = \binom{n+k-1}{k} = \binom{n+k-1}{n-1}.$$

Idea There are  $k-1$  additional choices because repetition is allowed.

$$* P(\text{event}) = \frac{\text{No. of options satisfying the event}}{\text{Total no. of options}}.$$

Ex4 You want to partition a line with

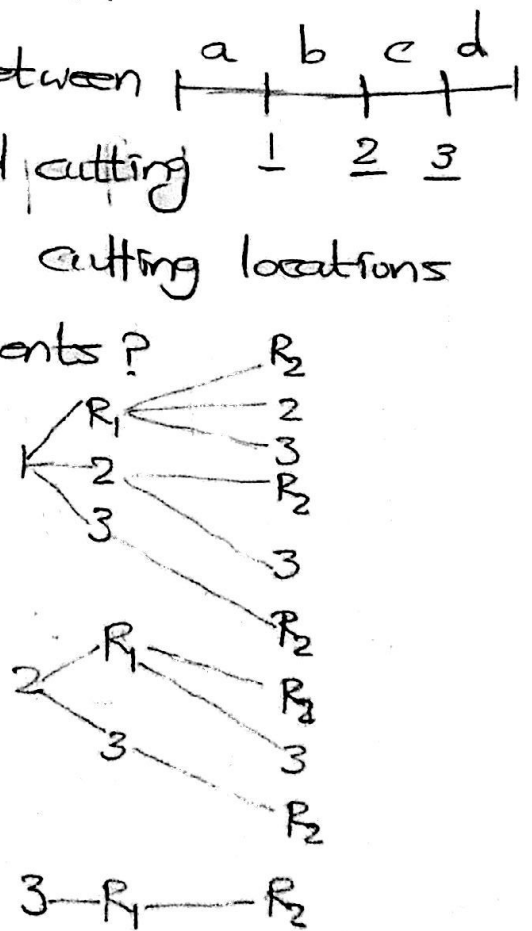
4 segments into a number of parts between 2 and 4 by picking 3 of the specified cutting locations. Segments are labeled, and cutting locations can be repeated. How many arrangements?

Sol. Here,  $n=3$  and  $k=3$ .

We sample with replacement, no ordering. We get 2 new options:

$$\{1, 2, 3, R_1, R_2\}.$$

$$\text{Ans.} = \binom{n+k-1}{k} = \binom{5}{3} = 10.$$



Ex 5 There are  $n$  boxes labeled  $1, 2, \dots, n$ . There are also  $n$  balls labeled  $1, 2, \dots, n$ . Consider the case of  $n \geq 2$ . You place the balls into the boxes. Each ball has an equal probability of being placed at any box.

Find the probability that after placing all balls 2

(a) Box 1 is (and possibly others are) empty.

(b) Exactly one box is empty.

Sol. No. of all possible arrangements =  $n^n$ .

(a) Probability is  $\frac{(n-1)^n}{n^n}$ .

(b) Probability is  $\frac{n(n-1) \binom{n}{2} (n-2)!}{n^n} = \frac{\binom{n}{2} n!}{n^n}$ .

[Choose empty box out of  $n$  —

choose box with 2 balls out of  $n-1$  —

choose 2 balls out of  $n$  for that box —

Arrange the rest a ball per box].