

## Review of Probability Theory

Randomness - the essence of communication is randomness.

↳ If A knew what B would say in advance, there'd be no need to say it! →

<sup>Communication</sup>  
- Assumes that transmitter is connected to a random source whose output the receiver cannot predict with certainty.

Random Processes - provide good models for info source (some probabilities associated w/ information event!)  
- noise ~ imperfections in communication that cause received signal to be different from transmitted.

1. <sup>deterministic</sup> linear distortion (eg.  $x(t) * h(t)$ )

2. non-deterministic Noise - modeled as random process!  
(random)

Probability Measure P - a set function that assigns non-negative values to events in  $\Omega$  sample space.

$$(1) 0 \leq p(E) \leq 1 \quad \forall E \in \Omega$$

$$(2) p(E_1 \cup E_2) = p(E_1) + p(E_2) - p(E_1 \cap E_2) \text{ generally.}$$

Certainly, if A, B are disjoint (mutually exclusive), i.e.  $A \cap B = \emptyset$ ,

$$\text{then } p(E_1 \cup E_2) = p(E_1) + p(E_2).$$

In general, if  $E_1, E_2, E_3, \dots$  are disjoint (i.e.  $E_i \cap E_j = \emptyset \quad \forall i \neq j$ )

$$\text{then } p\left(\bigcup_i E_i\right) = \sum_i p(E_i)$$

$$(3) p(\Omega) = 1, \quad p(\emptyset) = 0$$

$$(4) \text{ If } E_1 \subset E_2, \text{ then } p(E_1) \leq p(E_2)$$

Conditional Probability:

Consider events  $E_1$  and  $E_2$  with corresponding probabilities  $P(E_1)$  and  $P(E_2)$ .

If we receive info that event  $E_2$  has in fact happened, our probability about event  $E_1$  will no longer be  $P(E_1)$ .

We define the conditional probability of event  $E_1$ , given that event  $E_2$  has occurred as

$$P(E_1|E_2) = \frac{P(E_1 \cap E_2)}{P(E_2)}, \quad P(E_2) \neq 0.$$

(eg)  $\begin{cases} E_1: \text{It's snowing in Canada} \\ E_2: \text{It's winter} \end{cases}$   
 $P(E_1) \approx 1/4$   
 $P(E_2) \approx 1/3$

If  $P(E_1|E_2) = P(E_1)$ , then knowledge of  $E_2$  does not change  $P(E_1)$ .

In this case,  $P(E_1 \cap E_2) = P(E_1) \cdot P(E_2)$

and the events are statistically independent!

Total Probability Thm:

$$\text{If } \bigcup_{i=1}^n E_i = \Omega \quad \text{and} \quad E_i \cap E_j = \emptyset \quad \forall i \neq j, 1 \leq i, j \leq n$$

(collectively exhaustive)      (mutually exclusive or disjoint)

ie. events  $E_i, E_j$  partition the sample space ...

If we have the conditional probabilities of event  $A$  (ie.  $P(A|E_i)$  ~ the  $i$ th conditional probability)

then

$$P(A) = \sum_{i=1}^n P(A|E_i) \cdot P(E_i)$$

Using Bayes' rule, we can obtain

$$P(E_i|A) = \frac{P(A|E_i) \cdot P(E_i)}{P(A)} = \frac{P(A|E_i) \cdot P(E_i)}{\sum_{i=1}^n P(A|E_i) \cdot P(E_i)}$$

### Example: The Binary Symmetric Channel (BSC)

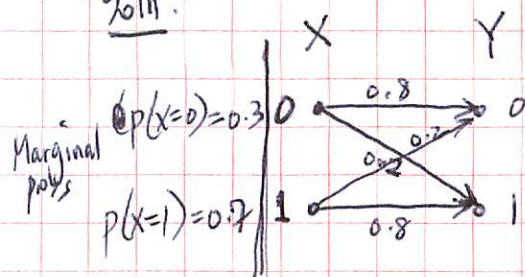
An information source produces 0 and 1 with probabilities 0.3 and 0.7 respectively. The output of the source is transmitted via a channel with crossover probability equal to 0.2. (probability of error)

What is the probability that

(a) A 1 is observed?

(b) A 1 was transmitted given that a 1 was observed at the channel output?

Soln:



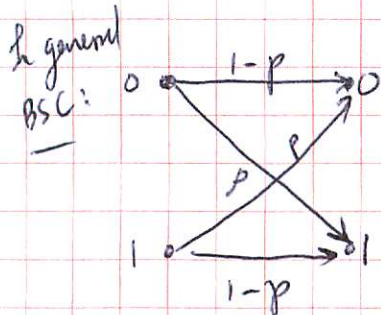
(a)  $p(y=1) = ?$

By total prob thm,

$$p(y=1) = p(y=1|x=0) \cdot p(x=0) + p(y=1|x=1) \cdot p(x=1)$$

$$= (0.2)(0.3) + (0.8)(0.7)$$

$$= 0.06 + 0.56 = \boxed{0.62}$$



(b)  $p(x=1|y=1) = ?$

By Bayes' thm,  $p(x=1|y=1) = \frac{p(y=1|x=1) \cdot p(x=1)}{p(y=1)}$

$$= \frac{(0.8)(0.7)}{(0.62) \sim \text{from above}}$$

$$= 0.56/0.62$$

$$= \boxed{0.903} \text{ - which is reasonable!}$$

for this channel