

Exam Review

ECE 313

Probability with Engineering Applications

Lecture 25

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Final Project Timeline

- **Project Summary: Saturday, Nov. 16, 5:00pm**
 - Submit a one-paragraph summary of your goals and overall plans
- **Short Progress Reports: Every Tuesday, starting Nov. 19.**
 - A short paragraph (3-4 lines) on what progress you made and what is the plan for the next week
- **First Presentation, this Thursday, Nov. 21, in the class**
 - 2 slides showing your overall plan, timeline, the division of tasks, and techniques are your going to use.
- **Intermediate Presentation, Tuesday, Dec. 3, in the class**
 - Reporting on your progress
- **Final Presentation, Saturday, Dec. 14, in the class**
- **Final Exam Review Session, Monday, Dec. 16, Time: TBD**
- **Final Project Report, December 19, 5:00pm.**

Review: Basic Concepts

- Basic Concepts:
 - **Random experiment** is an experiment the outcome of which is not certain
 - **Sample Space (S)** is the totality of the possible outcomes of a random experiment
 - **Discrete (countable) sample space** is a sample space which is either
 - *finite*, i.e., the set of all possible outcomes of the experiment is finite
 - *countably infinite*, i.e., the set of all outcomes can be put into a one-to-one correspondence with the natural numbers
 - **Continuous sample space** is a sample space for which all elements constitute a continuum, such as all the points on a line, all the points in a plane
 - An **event** is a collection of certain sample points, i.e., a subset of the sample space
 - **Universal event** is the entire sample space S
 - **The null set \emptyset is a null or impossible event**

Review: Algebra of Events

- **Algebra of Events**

- The **intersection** of E_1 and E_2 is given by:

- $E_1 \cap E_2 = \{s \in S \mid s \text{ is an element of both } E_1 \text{ and } E_2\}$

- The **union** E_1 and E_2 is given by:

- $E_1 \cup E_2 = \{s \in S \mid \text{either } s \in E_1 \text{ or } s \in E_2 \text{ or both}\}$

- In general: $|E_1 \cup E_2| \leq |E_1| + |E_2|$

- where $|A|$ = the number of elements in the set (**Cardinality**)

- Definition of *union* and *intersection* extend to any finite number of sets:

$$\bigcup_{i=1}^n E_i = E_1 \cup E_2 \cup E_3 \cup \dots \cup E_n$$

$$\bigcap_{i=1}^n E_i = E_1 \cap E_2 \cap E_3 \cap \dots \cap E_n$$

Review: Mutually Exclusive Events

- **Mutually exclusive or disjoint events** are two events for which

$$A \cap B = \emptyset$$

- A list of events A_1, A_2, \dots, A_n is said to be
 - composed of **mutually exclusive events** iff:

$$A_i \cap A_j = \begin{cases} A_i & \text{if } i = j \\ \emptyset & \text{otherwise} \end{cases}$$

- **collectively exhaustive** iff: $A_1 \cup A_2 \cup \dots \cup A_n = S$

Review: Probability Axioms

- **Probability Axioms**

- Let S be a sample space of a random experiment and $P(A)$ be the probability of the event A
- The probability function $P(\cdot)$ must satisfy the three following axioms:
- **(A1)** For any event A , $P(A) \geq 0$
(probabilities are nonnegative real numbers)
- **(A2)** $P(S) = 1$
(probability of a certain event, an event that must happen is equal 1)
- **(A3)** $P(A \cup B) = P(A) + P(B)$, whenever A and B are mutually exclusive events, i.e., $A \cap B = \emptyset$
(probability function must be additive)
- **(A3')** For any countable sequence of events $A_1, A_2, \dots, A_n, \dots$, that are mutually exclusive (that is $A_j \cap A_k = \emptyset$ whenever $j \neq k$)

$$P\left(\bigcup_{n=1}^{\infty} A_n\right) = \sum_{n=1}^{\infty} P(A_n)$$

Review: Probability Axioms

- **(Ra)** For any event A , $P(\overline{A}) = 1 - P(A)$
- **(Rb)** If \emptyset is the impossible event, then $P(\emptyset) = 0$
- **(Rc)** If A and B are any events, not necessarily mutually exclusive, then

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

- **(Rd)**(generalization of Rc) If A_1, A_2, \dots, A_n are any events, then

$$\begin{aligned} P\left(\bigcup_{i=1}^n A_i\right) &= P(A_1 \cup A_2 \cup \dots \cup A_n) = \sum_i P(A_i) - \sum_{1 \leq i < j \leq n} P(A_i \cap A_j) \\ &+ \sum_{1 \leq i < j < k \leq n} P(A_i \cap A_j \cap A_k) + \dots + (-1)^{n-1} P(A_1 \cap A_2 \cap \dots \cap A_n) \end{aligned}$$

where the successive sums are over all possible events, pairs of events, triples of events, and so on.

(Can prove this relation by induction (see class web site))

Review: Combinatorial Problems

- **Combinatorial Problems**

- Permutations with replacement:

- Ordered samples of size k , with replacement $P(n, k)$

- Permutations without replacement

- Ordered Samples of size k , without replacement

$$n(n-1)\dots(n-k+1) = \frac{n!}{(n-k)!} \quad k = 1, 2, \dots, n$$

- Combinations

- Unordered sample of size k , without replacement

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

- **Binomial Theorem**

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^k y^{n-k}$$

Review: Conditional Probability

- **Conditional Probability** of A given B ($P(A|B)$) defines the conditional probability of the event A given that the event B occurs and is given by:

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

if $P(B) \neq 0$ and is undefined otherwise.

- A rearrangement of the above definition gives the following ***multiplication rule (MR)***

$$P(A \cap B) = \begin{cases} P(B)P(A|B) & \text{if } P(B) \neq 0 \\ P(A)P(B|A) & \text{if } P(A) \neq 0 \\ 0 & \text{otherwise} \end{cases}$$

- Or:

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

Review: Bayes Formula

- **Theorem of Total Probability**
- Any event A can be partitioned into two disjoint subsets:

$$A = (A \cap B) \cup (A \cap \overline{B})$$

- Then:

$$\begin{aligned} P(A) &= P(A \cap B) \cup P(A \cap \overline{B}) \\ &= P(A | B)P(B) + P(A | \overline{B})P(\overline{B}) \end{aligned}$$

- In general:

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

- **Bayes Formula:**

$$P(B_j | A) = \frac{P(B_j \cap A)}{P(A)} = \frac{P(A | B_j)P(B_j)}{\sum_i P(A | B_i)P(B_i)}$$

Review: Independence of Events

- **Independence of Events:**
- Two events A and B are independent if and only if:

$$P(A|B)=P(A)$$

- Or events A and B are said to be independent if:

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

Review: Series and Parallel Systems

- **Reliability Applications:**

- Recovery blocks
- Series and parallel systems:

- Series System: $R_s = P$ (“The system is functioning properly.”)

$$= P(A_1 \cap A_2 \cdots \cap A_n)$$

$$= P(A_1)P(A_2) \cdots P(A_n)$$

$$= \prod_{i=1}^n R_i \quad (2.1)$$

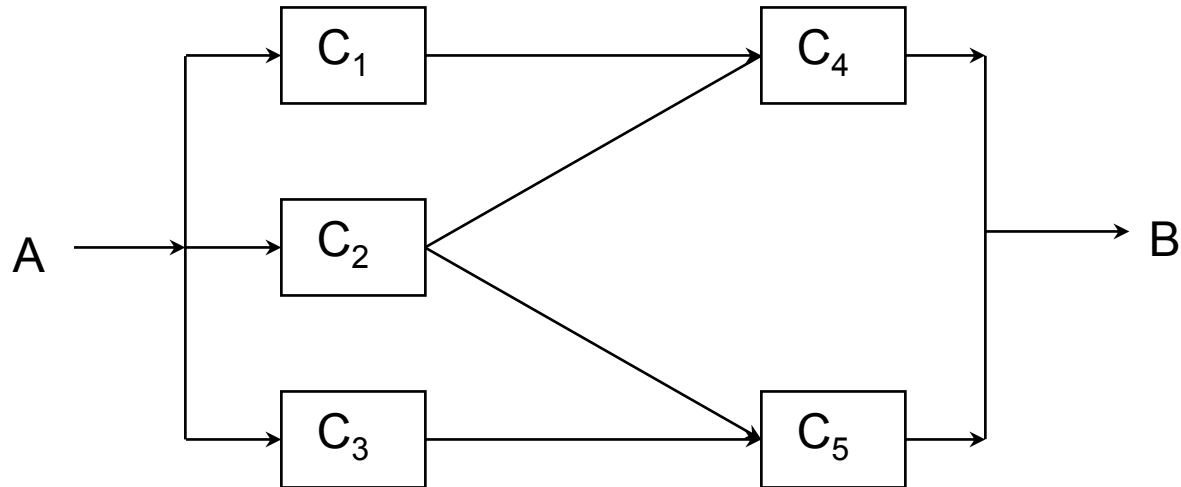
- Parallel System:

$$R_p = 1 - F_p = 1 - \prod_{i=1}^n (1 - R_i)$$

- In general: $R_{sp} = \prod_{i=1}^n [1 - (1 - R_i)^{n_i}]$

Review: Complex Series-Parallel Systems

- Non series-parallel systems:



- Using the theorem of total probability and the Bayes formula:

$$P(X) = P(X | X_2)P(X_2) + P(X | \bar{X}_2)P(\bar{X}_2) = P(X | X_2)R_2 + P(X | \bar{X}_2)(1 - R_2)$$

Review: TMR System

- Bernoulli Trials**

- The probability of obtaining exactly k successes in n trials is :

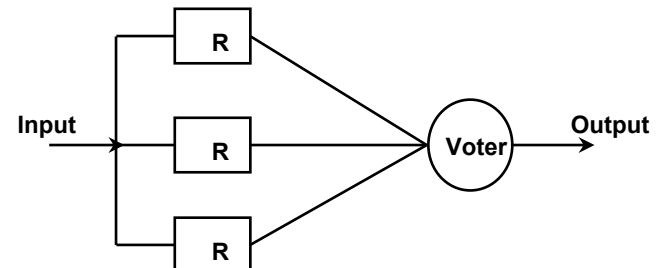
$$p(k) = \binom{n}{k} p^k q^{n-k} \quad k = 0, 1, \dots, n$$

- NMR System:**

$$\begin{aligned} R_{m|n} &= P(\text{"}m \text{ or more components functionin g properly "}) \\ &= P\left(\bigcup_{i=m}^n \{\text{" exactly } i \text{ components functionin g properly "}\}\right) \\ &= \sum_{i=m}^n P(\text{" exactly } i \text{ components functionin g properly "}) \\ &= \sum_{i=m}^n p(i) = \sum_{i=m}^n \binom{n}{i} R^i (1 - R)^{n-i} \end{aligned}$$

- TMR System:**

$$R_{TMR} = 3R^2 - 2R^3$$



Review: Random Variables

- **Random Variables:**

- A random variable X on a sample space S is a function $X: S \rightarrow \mathbb{R}$ that assigns a real number $X(s)$ to each sample point $s \in S$.
- ***Discrete*** random variables: The random variables which are either finite or countable.
 - Bernoulli
 - Binomial
 - Poisson
 - Geometric
 - Modified Geometric
- ***Continuous*** random variables: The random variables that take on a continuum of possible values.
 - Uniform
 - Normal
 - Exponential

Review: Cumulative distribution function (cdf)

- **Cumulative distribution function (cdf) (or distribution function)** $F(\cdot)$ of a random variable X is defined for any real number $b, -\infty < b < \infty$, by $F(b) = P\{X \leq b\}$
- $F(b)$ denotes the probability that the random variable X takes on a value that is less than or equal to b .
- Some properties of cdf F are:
 - i. $F(b)$ is a non-decreasing function of b ,
 - ii. $\lim_{b \rightarrow +\infty} F(b) = F(\infty) = 1$,
 - iii. $\lim_{b \rightarrow -\infty} F(b) = F(-\infty) = 0$.
- All probability questions about X can be answered in terms of cdf $F(\cdot)$. e.g.: $P\{a \leq X \leq b\} = F(b) - F(a)$ for all $a < b$

Review: Discrete Random Variables

- **Discrete Random Variables:**
 - **Probability mass function (pmf):**

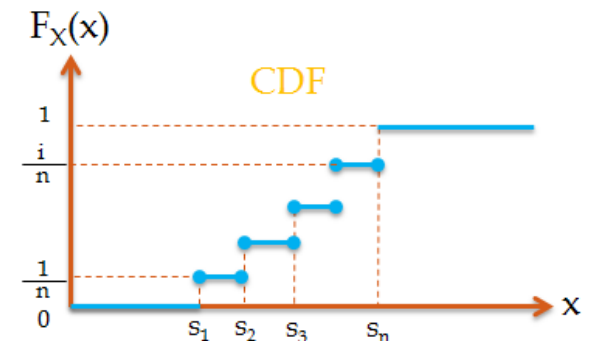
- Properties:
 - $p(a) = P\{X = a\}$
 - $$\begin{cases} p(x_i) > 0, & i = 1, 2, \dots \\ p(x) = 0, & \text{for other values of } x \end{cases}$$

$$\sum_{i=1}^{\infty} p(x_i) = 1$$

- **Cumulative distribution function (CDF):**

$$F(a) = \sum_{\text{all } x_i \leq a} p(x_i)$$

- **A stair step function**



Review: Continuous Random Variables

- **Continuous Random Variables:**
 - **Probability distribution function (pdf):**

$$P\{X \in B\} = \int_B f(x) dx$$

- Properties:

$$1 = P\{X \in (-\infty, \infty)\} = \int_{-\infty}^{\infty} f(x) dx$$

- All probability statements about X can be answered

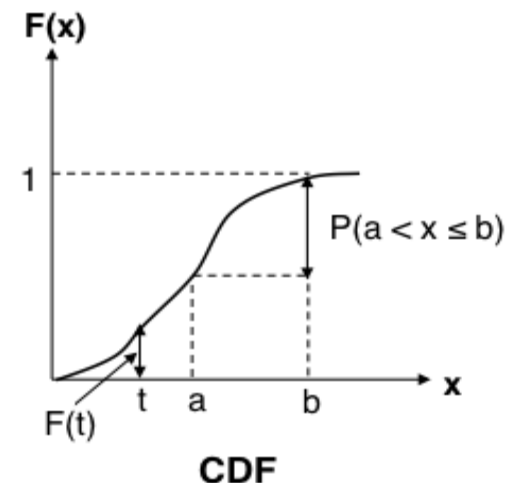
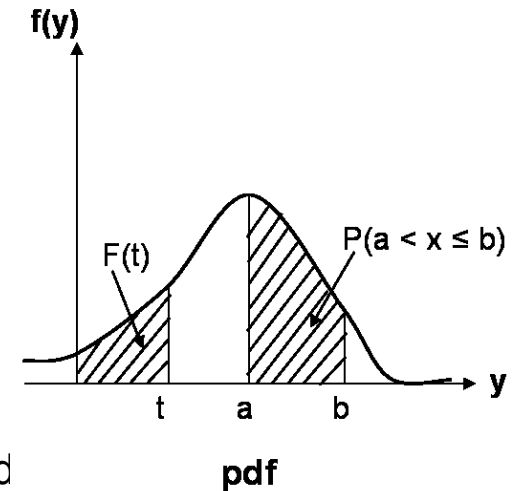
$$P\{a \leq X \leq b\} = \int_a^b f(x) dx$$

$$P\{X = a\} = \int_a^a f(x) dx = 0$$

- **Cumulative distribution function (CDF):**

$$F_x(x) = P(X \leq x) = \int_{-\infty}^x f_x(t) dt, \quad -\infty < x < \infty$$

- Properties: $\frac{d}{da} F(a) = f(a)$
- **A continuous function**



Review: Important Distributions

- Summary of important distributions:

Distribution	PDF or PMF	Mean	Variance
$Bernoulli(p)$	$\begin{cases} p, & \text{if } x = 1 \\ 1 - p, & \text{if } x = 0. \end{cases}$	p	$p(1 - p)$
$Binomial(n, p)$	$\binom{n}{k} p^k (1 - p)^{n-k}$ for $0 \leq k \leq n$	np	npq
$Geometric(p)$	$p(1 - p)^{k-1}$ for $k = 1, 2, \dots$	$\frac{1}{p}$	$\frac{1-p}{p^2}$
$Poisson(\lambda)$	$e^{-\lambda} \lambda^x / x!$ for $k = 1, 2, \dots$	λ	λ
$Uniform(a, b)$	$\frac{1}{b-a} \forall x \in (a, b)$	$\frac{a+b}{2}$	$\frac{(b-a)^2}{12}$
$Gaussian(\mu, \sigma^2)$	$\frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$	μ	σ^2
$Exponential(\lambda)$	$\lambda e^{-\lambda x} \quad x \geq 0, \lambda > 0$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$