

**Department of Systems Engineering**  
**George Mason University**

**SYST 302: Systems Methodology  
and Design II #3**

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# **Probability Concepts**

- **Concepts and Theory**
- **Probability Distribution and Random Variables**
- **Probability Inference**

# Probability Concept

- **Motivation**

- Decision making under uncertainty (risk)
- Need probability model to quantitatively evaluate different criteria
- Rely on statistical regularity to predict future behavior

- **Concept**

- Probability is usually associated with an *event*
- An *event* is a subset of a *sample space*
- A *sample space* is a set of all possible *outcomes*
- An *outcome* is the result of a *random experiment*
- A *random experiment* is a process by which a observation is obtained (ex: throwing a die)

# Fundamentals

- **Definitions**

- **Given a sample space  $S$ , for each event  $A$ , there is a number  $P(A)$  called probability of  $A$ , such that the following axioms of probability are satisfied**

- $0 \leq P(A) \leq 1$

$$P(S) = 1$$

$$P(A \cup B) = P(A) + P(B) \text{ if } A \cap B = \emptyset$$

Thm:  $P(A^c) = 1 - P(A)$  and  $P(S) = P(A) + P(A^c) = 1$

where  $A^c = S \setminus A$

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

# Conditional Probability

- **Conditional Probability  $P(B|A)$** 
  - The probability of an event B under the condition that an event A occurs

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

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

$$\text{Bayes Rule : } P(B | A) = \frac{P(A \cap B)}{P(A)} = \frac{P(A | B)P(B)}{P(A)}$$

- **Independent Events**

*if  $P(A \cap B) = P(A)P(B)$  then A and B are independent*

*$\Rightarrow P(A|B) = P(A)$  or  $P(B|A) = P(B)$*

# Random Variables

- **Random Variable**
  - A variable  $x$ , about which we are uncertain, and to which we can assign a real number
  - Defined on the sample space
- **Discrete Random Variable**
  - Can assume finite number of discrete values
  - Probability is well defined for each value
  - Described by probability mass function
- **Continuous Random Variable**
  - Can assume any real value in a specified range
  - Probability is well defined for any interval
  - Described by probability density function

# Probability Distribution

- **Probability Mass Function (pmf)**

- Define the probability that a random variable  $x$  takes on a value

$$f(x) = \begin{cases} P_j & \text{if } x = x_j \\ 0 & \text{otherwise} \end{cases}$$

$$\text{CDF: } F_x(x) = P(x_j \leq x) = \sum_{x_j < x} f(x_j) = \sum_{x_j < x} P_j$$

- **Probability Density Function (pdf)**

- Define the probability that a random variable  $x$  takes on a value in an interval  $[a, b]$

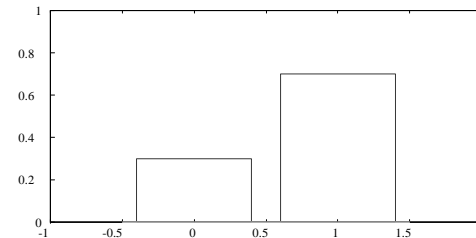
$$P(a \leq x \leq b) = \int_a^b f_x(x) dx \quad \text{where} \quad \int_{-\infty}^{\infty} f_x(x) dx = 1$$

$$\text{CDF: } F_x(x_0) = P(x \leq x_0) = \int_{-\infty}^{x_0} f_x(x) dx \quad \text{Note that } P(x > x_0) = 1 - F_x(x_0)$$

# Discrete Random Variables

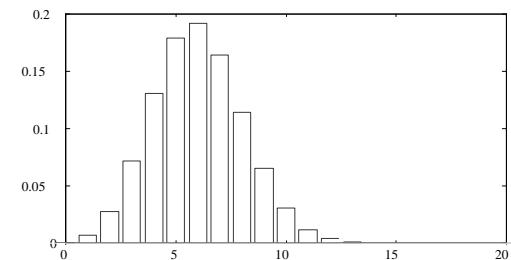
(1) *Bernoulli R.V.*  $S_x = \{0, 1\}$

$$P_0 = P, P_1 = 1 - P$$



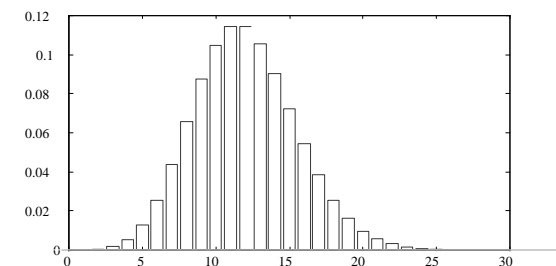
(2) *Binomial R.V.*  $S_x = \{0, 1, 2, \dots, n\}$

$$P_k = \frac{n!}{(n-k)!k!} P^k (1-P)^{n-k}, \quad k = 0, 1, 2, \dots, n$$



(3) *Poisson R.V.*  $S_x = \{0, 1, 2, \dots\}$

$$P_k = \frac{\alpha^k}{k!} e^{-\alpha}, \quad k = 0, 1, 2, \dots \text{ and } \alpha > 0$$





# Examples

## (1) *Binomial*

The probability of having at least two defects in 10 samples when the defect probability is 0.01

$$P_{\geq 2} = 1 - P_0 - P_1 = 1 - (0.99)^{10} - \frac{10!}{9!1!} (0.01)^1 (0.99)^9 \approx 0.043$$

## (2) *Poisson*

If on the average, 3 cars enter a parking lot per minute, what is the probability that during any minute 4 or more cars will enter the lot?

$$P_{\geq 4} = 1 - P_0 - P_1 - P_2 - P_3 = 1 - e^{-3} \left( \frac{3^0}{0!} + \frac{3^1}{1!} + \frac{3^2}{2!} + \frac{3^3}{3!} \right) \approx 0.353$$

# Continuous Random Variables

(1) *Uniform R.V.*  $S_x = [a, b]$

$$f(x) = 1/(b - a), a \leq x \leq b$$

(2) *Exponential R.V.*  $S_x = [0, \infty)$

$$f(x) = \lambda e^{-\lambda x}, x \geq 0 \text{ and } \lambda > 0$$

(3) *Gaussian R.V.*  $S_x = (-\infty, \infty)$

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-m)^2}{2\sigma^2}}, -\infty < x < \infty \text{ and } \sigma > 0$$

# Examples

## (1) *Exponential*

The average life time of a system is 4 hours.

What is the probability that a system will fail within the first two hours of operation?

$$P_{\leq 2} = \int_0^2 \frac{1}{4} e^{-\frac{t}{4}} dt \approx 0.39$$

## (2) *Gaussian*

If on the average, a product weighs 10 kg with STD = 1kg, what is the probability that the product is lighter than 9 kg?

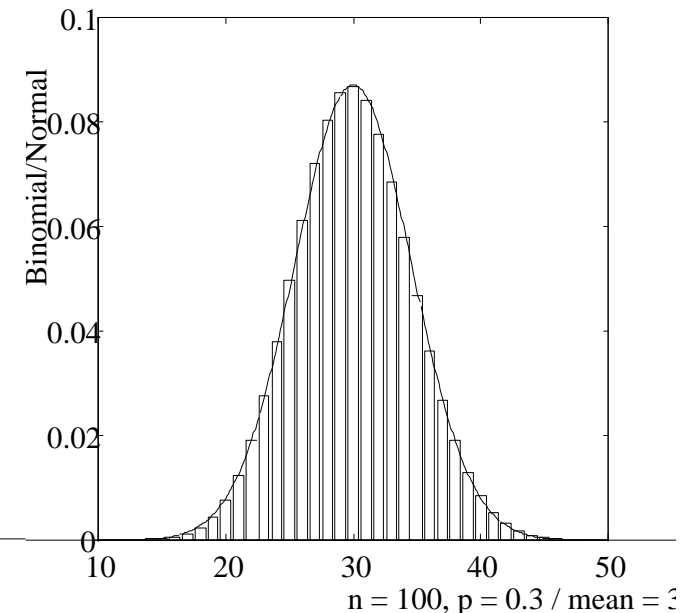
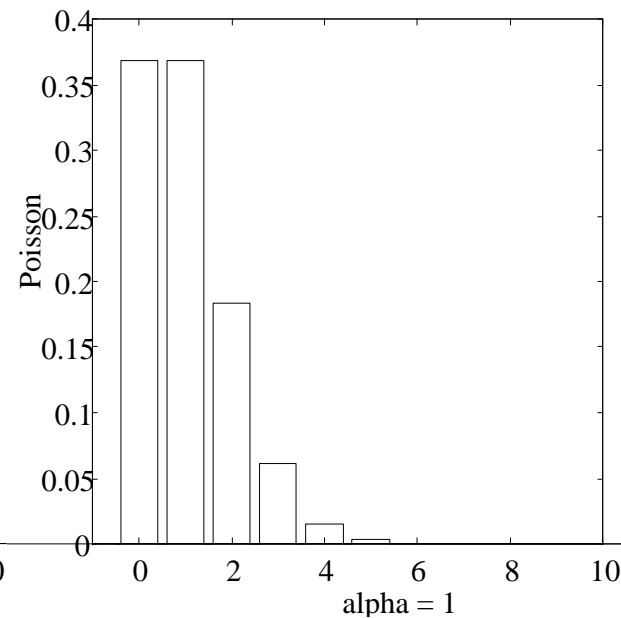
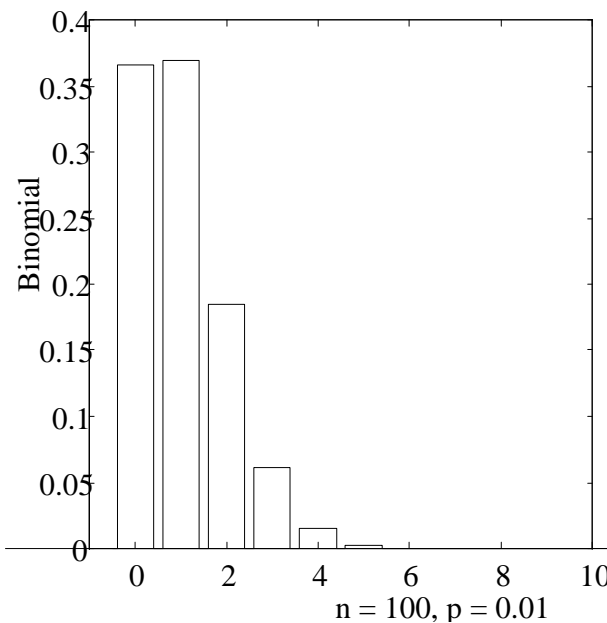
$$P_{\leq 9} = \int_{-\infty}^9 \frac{1}{\sqrt{2\pi}} e^{-\frac{(t-10)^2}{2}} dt \approx 0.159$$

# Expectation and Variance

- **Expectation (Mean)**  $\bar{x} = E(x) = \int_{-\infty}^{\infty} x f_x(x) dx$
- **Variance**  $\sigma_x^2 = E[(x - \bar{x})^2] = \int_{-\infty}^{\infty} (x - E[x])^2 f(x) dx = E[x^2] - E[x]^2$

	pdf	mean	variance
<i>Bernoulli</i>	$P_0 = 1 - p, P_1 = p$	$p$	$p(1 - p)$
<i>Binomial</i>	$\frac{n!}{(n-k)!k!} P^k q^{n-k}$	$np$	$npq$
<i>Poisson</i>	$\frac{\alpha^k}{k!} e^{-\alpha}$	$\alpha$	$\alpha$
<i>Uniform</i>	$1/(b-a)$	$(a+b)/2$	$(b-a)^2/12$
<i>Exponential</i>	$\lambda e^{-\lambda x}$	$1/\lambda$	$1/\lambda^2$
<i>Normal</i>	$\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-m)^2}{2\sigma^2}}$	$m$	$\sigma^2$

# Binomial and Poisson/Normal



$$P_k = \frac{n!}{(n-k)!k!} P^k (1-P)^{n-k}$$

$$P_k = \frac{\alpha^k}{k!} e^{-\alpha}$$

$$m = 30, \sigma^2 = 21$$

# Sample Mean and Variance

$$\text{Sample Mean: } \bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n} = \frac{\sum_i x_i}{n}$$

$$\text{Sample Variance*}: \sigma^2 = \frac{\sum_i (x_i - \bar{x})^2}{n-1} = \frac{\sum_i x_i^2 - \left(\sum_i x_i\right)^2 / n}{n-1}$$

\* Unbiased estimate

# Sum of Random Variables and Central Limit Theorem

Let  $S_n = x_1 + x_2 + \dots + x_n$

where  $x_1, x_2, \dots, x_n$  are i.i.d. with mean  $\mu$  and variance  $\sigma^2$ ,  
then

$$\lim_{n \rightarrow \infty} f_{S_n}(s) = N(n\mu, n\sigma^2)$$

or

$$\lim_{n \rightarrow \infty} f_{Z_n}(z) = N(0, 1) \quad \text{where } Z_n = \frac{S_n - n\mu}{\sqrt{n}\sigma}$$

# Monte Carlo Simulation

- **Motivation**

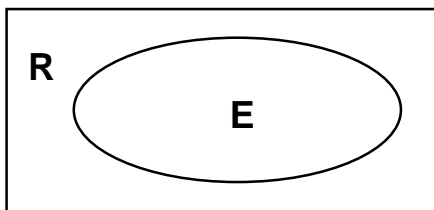
- Uncertain decision environment requires probabilistic models
- Formal mathematic solutions may be difficult or impossible
- Monte Carlo analysis provides a powerful tool through simulation

- **Applications**

- System performance evaluation
- System operation characteristics analysis

- **A Simple Example**

- Determine the area of an irregular shape



(1) Generate 2 - D points in R uniformly

(2)  $\text{Area}(E) \approx \frac{\text{\# of points in } E}{\text{total \# of points}} \text{Area}(R)$