

1) What is Simulation? When Simulation is the appropriate tool and when it is not appropriate?

→ A simulation is the limitation of the operation of a real world process or system over times.

1) When simulation is the appropriate tool:

The availability of special-purpose simulation language of massive computing capabilities at a discussing operation, Simulation can be used for the following purposes.

a) Simulation enables the study of and experimentation with the internal interactions of a complex system.

b) Informational, organizational and environmental changes can be simulated, and the effect of these attractions on the models behaviour can be observed. Simulation can be used as a pedagogical device to reinforce analytic solution methodologies.

2) When simulation is not Appropriate:

a) Simulation should not be used when the problem can be solved by common sense.

b) Simulation should not be used if the problem can be solved analytically.

c) Simulation should not be used if it is less expensive to perform direct experiments.

d) Not to use simulation if the cost exceeds the savings.

2) Advantages and disadvantages of simulation.

→ a) New policies, operating procedures, decision rules, information flows, organizational procedures and so on can be explored without disrupting ongoing operations of the real system.

b) New hardware designs, physical layouts, transportation systems and so on can be used without committing resources for their acquisition.

- 2
- ③ hypothesis about how or why certain phenomena occur can be tested for feasibility.
 - ④ Insight can be obtained about the interaction of variables.
 - ⑤ Time can be compressed or expanded to allow for a speed-up or slow-down the phenomena under investigation.

Disadvantages:

- ① Model Building requires special training. It is - an art that is learned overtime and through experience. Furthermore, if two models are connected by different competent individuals, they might have similarities, but it is highly unlikely that they will be the same.
- ② simulation results can be difficult to interpret. Most simulation outputs are eventually random variables. So it can be hard to distinguish whether an observation is the result of system inter relations.
- ③ Simulation modelling an analysis can be time consuming and expensive. Skimping on resources for modelling and analysis could result in a simulation model or analysis that is not sufficient to the task.

3) Area of Applications.

→ The application of simulation are vast. Some are explained below.

④ Manufacturing Application:

- Methodology for selecting the most suitable bottleneck detection methods.
- Automating the development of stipyard manufacturing models.
- Facilitation in manufacturing engineering process

⑤ Water Fabrication:

- A paradigm shift is assigning lots to tools
- scheduling a multi-chip package assembly lens with central process

⑥ Business Processing:

- A new policies for the services request assignment problem.
- A process execution monitoring and adjustment schemes.

⑦ Construction Engineering and project management:

- Scheduling of limited Bar-Binders over multiple Buildings.
- Process execution, monitoring and adjustment schemes.

③ Logistics, Transportation and Distribution:

- operating policies for a Bag transportation System.
- Dispensing plan for emergency medical supplies in this event of bioterrorism.

④ Military Applications:

- Multinational Inter-theater logistics coordination.
- Training joint forces for asymmetric operations.

⑤ Health Care:

- Supporting smart thinking to improve hospital performance.
- Infection disease control policy.
- Reducing emergency department overcrowding.

4) Systems and System Environment-

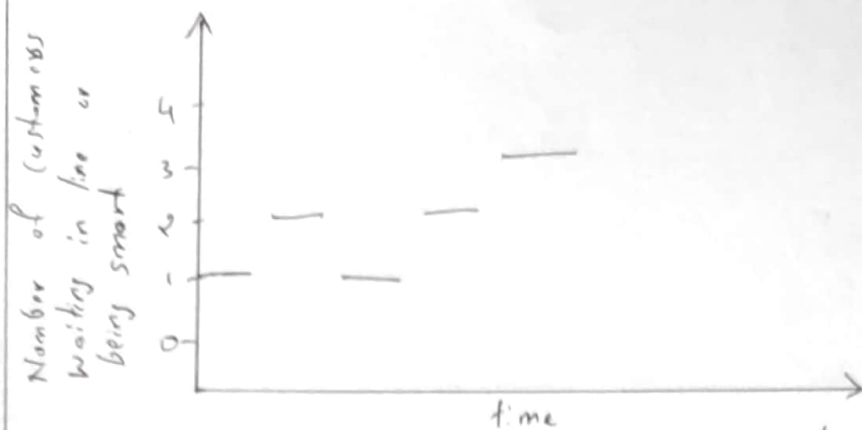
- A System is defined as a group of objects that are joined together in some regular instructions or independence toward the accomplishment of some purpose. A system is often affected by changes occurring outside the system. Such changes are said to occur in the system environment. In modelling systems, it is necessary to decide on the boundary between the system and its environment.

5) Components of a system.

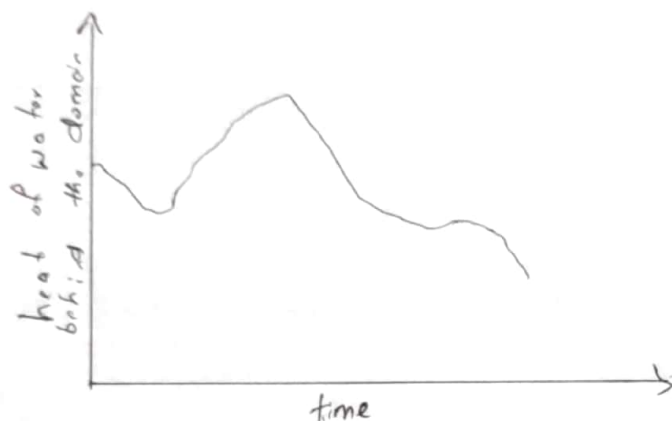
- An entity is an object of interest in the system. An attribute is a property of an entity. An activity represents a time period of specified length. The state of a system defined to be that collection of variables necessary to describe the system at any simulation to the objectives of the study. An event is defined as an instructions occurrence that might change the state of the system.

6) Discrete and Continuous Systems.

- A discrete system is one in which the state variable changes only at a discrete set of points in time.



A continuous system is one which the state variables change continuously over time.



7) Model of a system

A model is defined as a representation of a system for the purpose of studying that system. For more studies it is only necessary to consider those aspects of the system that affect the problem under investigation. These aspects are represented in a model of the system, the model by definition is a simplification of the system. On the other hand, the model should be sufficiently detailed to permit valid conclusions to be drawn about the real system. Different models of the same system could be required as the purpose of investigation changes.

8) Types of models: models can be classified as being mathematical or physical. A mathematical model uses symbolic notation and mathematical equations to represent a system. A physical model is a large or small version of an object such as the enlargement of an atom or a scaled-down of the solar system. A static simulation model sometimes called as monte-carlo simulation, represents a system at a particular time. Dynamic simulation model represents system as they change over time. Simulation models that contain no random variables are classified as deterministic.

5 9) Discrete - Event System simulation

→ Discrete - event system simulation is the modelling of systems in which the state variable changes only at a discrete set of points in time. This simulation models are analyzed by numerical rather than analytical methods. Analytical methods employ the deductive reasoning of mathematics to solve the model. Real world simulation models are rather large and the amount of data stored and manipulated is vast, so such runs are usually conducted with the aid of a computer.

10) Steps in a simulation study.

→ ① Problem Formulation:

In many instances, policy makers and analysts are aware that there is a problem long before the nature of the problem is known.

② Setting of objectives and overall project plan:

The objectives indicate the questions to be answered by simulation. At this point, a determination should be made concerning whether simulation is the appropriate methodology for the problem as formulated and the objectives as stated.

③ Model Conceptualization:

The construction of a model of a system is probably as much an art as science. The art of modeling is enhanced by an ability to abstract the essential features of a problem to select and modify basic assumptions that characterize the system and then to test and elaborate the model until approximation results.

④ Data collection:

There is a constant interplay between the construction of the model and the collection of the needed input data. As the complexity of the model changes, the required data elements can also change.

⑤ Model translation:

More real world systems result in models that require a great deal of information storage and compilation. So the model must be entered into a computer recognizable format.

⑥ Verification:

Verification pertains to the complete program that has been prepared for the simulation model. The computer program performing properly with computer models.

⑦ Validated:

Validation usually is achieved through the calibration of the model on iterative process of comparing the model against actual system behaviour and using the discrepancy between the two, and the insights gained to improve model.

⑧ Experimental design:

The alternatives that are to be simulated must be determined. Often the decision concerning which alternatives to simulate will be a function of runs that has been completed and Analyzed.

⑨ Production runs and analysis:

Production runs and this subsequent analysis, are used to estimate measure of performance for the system design that are being simulated.

⑩ More runs:

Given the analysis of runs have been completed, the analyst determines whether additional runs are needed and what design these additional experiment should follow.

⑪ Documentation and Reporting:

There are 2 types of documentation program and progress. If the program is going to be used again by the same or different analysts, it could be necessary to understand how the program operates. This will create a confidence in the program. So that model users and policies makers can make decisions based on the analysis. Also, if the program is to be modified by the same or a different analyst, this step can be greatly facilitated by adequate documentation.

⑫ Implementation:

The success of implementation phase depends on how well the previous process step have been performed. It is also confident upon how thoroughly the analyst has investigated the document ultimate model uses during the entire simulation process.

11) Preview of terminology and concepts

→ ① Discrete random variables:

Let x be a random variable of the number of possible values of x is finite or countably infinite and is called discrete random variables. The possible values of x may be listed as x_1, x_2, \dots, x_m etc.

② Continuous random variables:

if the range space R_X of the random variable X is an interval or a collection of intervals. X is called a continuous random variable. For continuous random X , the probability that X lies in interval $[a, b]$ is given by

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

The function $f(x)$ is called probability density function of the random variable X . The probability density function satisfies the following conditions.

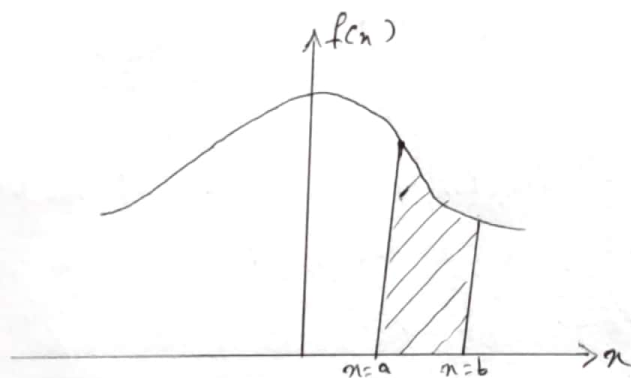
a. $f(x) \geq 0$ for all x in R_X

b. $\int_{-\infty}^{\infty} f(x) dx = 1$

c. $f(x) = 0$ if x is not in R_X

As a result of equation ①, for any specified value x_0 , $P(X=x_0) = 0$ because

$$\int_{x_0}^{x_0} f(x) dx = 0$$



$P(X=x_0) = 0$ also means that the following equation hold

$$P(a \leq X \leq b) = P(a < X < b) = P(0 \leq X < b) = P(a < X < b)$$

③ Cumulative distribution function:

The cumulative distribution function (cdf) denoted by $F(x)$ measures the probability that the random variable X assumes a value less than or equal to x that is,

$$F(x) = P(X \leq x)$$

$$F(x) = \sum_{x_i} P(x_i)$$

if X is continuous then

$$F(x) = \int_{-\infty}^x f(t) dt$$

All the probability questions about X can be answered in terms of

$$P(a < X \leq b) = F(b) - F(a) \text{ for all } a < b$$

④ Expectation :

An important concept in probability theory is that of the expectation of a random variable of x is a random variable. The expected value of x denoted by $E(x)$ for discrete and continuous variable is defined as follows.

$$E(x) = \sum x_i P(x_i) \text{ if } x \text{ is discrete.}$$

and

$$E(x) = \int_{-\infty}^{\infty} x f(x) dx \text{ if } x \text{ is continuous}$$

The expected value $E(x)$ of a random variable x is also referred to as the mean, is or the first moment of x . The quantity $E(x^n)$, $n \geq 1$ is called the n th moment of x and is computed as :

$$E(x^n) = \sum x_i^n P(x_i) \text{ if } x \text{ is discrete.}$$

and

$$E(x^n) = \int_{-\infty}^{\infty} x^n f(x) dx \text{ if } x \text{ is continuous.}$$

⑤ The mode :

The mode is used in describing several statistical models in the discrete cases, the mode is the value of the random variable that occurs most frequently. In the continuous case, the mode is the value of which the pdf is maximised.

12) Useful Statistical models :

→ ① Queueing Systems :

In queueing examples, interval and service time patterns, the times between arrivals and the service times are often probabilistic. However, it is possible to have a constant interarrival time, or a constant service time.

② Inventory and Supply-chain System :

In reactive inventory and supply chain system, there are at least three random variables, the number of units demanded per order, or per time period. The time between demands and time.

③ Reliability and maintainability : Time to failure has been modelled with numerous distributions, including the exponential, gamma and weibull. If only random failure occurs, the time to failure distribution may be modelled as exponential. The gamma distribution arises from modelling stand by redundancy to have each component has an exponential time to failure.

9) Limited data:

In many instances, simulations begin before data collection has been completed. There are all those distributions that have application to incomplete or limited data. There are the uniform, triangular and beta distributions.

⑤ Other distributions:

Several other distributions may be useful in discrete system simulation. The Bernoulli and binomial distributions are two discrete distributions which might describe phenomena of interest. The hyper exponential distribution is similar to the exponential distribution, but its greater variability might make it useful in certain instances.

13) Discrete distributions

→ ① Bernoulli trials and the Bernoulli distribution:

Consider an experiment consisting of n trials, each of which can be a success or a failure. Let $x_j = 1$ if the j th experiment resulted in a success and let $x_j = 0$ if the j th experiment resulted in a failure. The n Bernoulli trials are called a Bernoulli process if the trials are independent, each trial has the two possible outcomes, and the probability of a success remains constant from trial to trial. Thus

$$P(x_1, x_2, x_3, \dots, x_n) = p_1(x_1) \cdot p_2(x_2) \cdot p_3(x_3) \dots p_n(x_n)$$

and

$$p_i(x_i) = p(x_j) = \begin{cases} p & x_j = 1, j = 1, 2, \dots, n \\ 1-p = q & x_j = 0, j = 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

② Binomial distribution:

The random variable x that denotes the number of success in n Bernoulli trials has a binomial distribution given by $p(x)$, where

$$p(x) = \begin{cases} \binom{n}{x} p^x q^{n-x}, & x = 0, 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

③ Geometric and Negative Binomial distributions:

The geometric distribution is related to a sequence of Bernoulli trials, the random variable of interest x , is defined to be the number of trials to achieve the first success. The distribution of x is given by

$$p(x) = \begin{cases} q^{x-1} p, & x = 1, 2, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

10) ④ Poisson Distribution:

The poisson distribution describes many random process quite well and is mathematically quite simple. The poisson distribution was introduced in 1837 by S-D poisson in a book concerning criminal & civil justice matters.

$$p(n) = \begin{cases} \frac{e^{-\lambda} \lambda^n}{n!} & , n = 0, 1, \dots \\ 0 & , \text{otherwise} \end{cases}$$

14) Continuous distribution:

→ ① Uniform distribution: A random variable x is uniformly distributed on the interval $[a, b]$ if its pdf is given by

$$f(x) = \begin{cases} \frac{1}{b-a} & , a \leq x \leq b \\ 0 & , \text{otherwise} \end{cases}$$

The pdf is given by

$$F(x) = \begin{cases} 0 & , x \leq a \\ \frac{x-a}{b-a} & , a \leq x \leq b \\ 1 & , x > b \end{cases}$$

② Exponential distribution:

A random variable x is said to be exponentially distributed with parameter $\lambda > 0$ if its pdf is given by

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & , x > 0 \\ 0 & , \text{otherwise} \end{cases}$$

③ Gamma distribution:

A function used in defining the gamma distribution is the gamma function, which is defined as $B > 0$ as

$$\Gamma(B) = \int_0^{\infty} x^{B-1} e^{-x} dx$$

15) Poisson process:

→ Consider random events such as the arrival of job shop, the arrival of email service the arrival of boats to a dock, the arrival of calls to a call center the breakdown of machine in a large factory, and so on. These events may be divided by a counting function $N(t)$ defined for all $t \geq 0$. This counting function will represent the number of counts that occurred in $(0, t)$

Empirical distributions

- 16) An empirical distribution which may be either discrete or continuous, is a distribution whose parameters are the observed values in a sample of data. This is in contrast to parameter distribution families, which are characterized by specifying using number of parameters. Such as the mean and variance. An empirical distribution may be used when it is impossible or unnecessary to establish that a random variable has any particular parameter. Distributions are advantageous of an empirical distribution is that nothing is assumed beyond the observed values in the sample.

17) Characteristics of queuing system.

- ① The calling population: The population of potential customers, referred to as the calling population may be assigned to the finite or infinite. In systems with a large number of potential customers the calling population is usually assumed to be finite.
- ② System Capacity: In many queuing systems, there is a limit to the number of customers that may be in the waiting line of system. An arriving customer who finds the system full does not enter but returns immediately to the calling position.
- ③ The arrival process: The arrival process for infinite population models is usually characterized of intervals of inter times of successive customers. Arrivals may occur at scheduled times or at random times. When at random times the interval times are usually characterized by a probability distribution.
- ④ Queue behaviour & Queue discipline: Queue behaviour refers to the actions of customers while in queue waiting for service to begin. There is a possibility that incoming customers will talk. Queue discipline refers to the logical order of customers in a queue and determines which customer will be chosen.

18) Queuing Notation.

- Recognizing the diversity of queuing systems, and all proposed a notational system in parallel server system which has been widely adopted. An abridged version of this convention is based on the format $A/B/c/K/k$. These letters represent the following system characteristics.
- A represents interarrival time distribution, B represents service time distribution, C represents number of parallel servers, K represents system capacity, k represents size of the calling population.

long-run measures of performance of queuing system.

The primary long-run measures of performance of queuing systems are the long-run time-average number of customers in the system, and in the queue. The long-run average time spent in system and in queue (w) per customer and the service utilization or proportion of time that server is busy (ρ). The term system overall refers to the waiting line plus the service mechanism in and can refer to any subsystem of the queuing system on the other hand, the term queue refers to the waiting line alone. Other measure of performance of interest it includes the long-run proportion of customers.

20) Steady-state behaviour of M/G/1 queue.

→ For any M/G/1 queue if lines are too long, they can be reduced by decreasing the service utilization or by decreasing the service time variability. These remarks hold for almost all queues not just the M/G/1 queue. The utilization factor ρ can be reduced by decreasing the arrival rate, by increasing the service rate or by increasing the number of servers, because in general $\rho = \lambda / (c\mu)$, where c is number of parallel servers.

a1) Network of Queue

→ Many systems are naturally modeled as networks of single queues in which customer proceeding into one queue may be routed to another. The following results assume a stable system with an infinite calling population and no limits as system capacity.

① Provided that no customers are created or destroyed in the queue, then the departure rate of a queue is the same as the arrival rate into the queue over the long run.

② If customers arrive to queue i at rate λ_i , and a fraction ρ_{ij} of them are routed to queue j upon departure, then the arrival rate from queue i to j is $\lambda_i \rho_{ij}$ over the long run.

③ The overall arrival rate into queue j , λ_j is the sum of the arrival rate from all sources. If customers arrive from outside the network at rate a_j , then

$$\lambda_j = a_j + \sum_{i=1}^n \lambda_i \rho_{ij}$$

④ if queue j has $C_j < \infty$ parallel servers, each working at rate μ then in the long run utilization of each server is,

$$\rho_j = \frac{\lambda_j}{C_j \mu_j}$$

22) Models, optimization via simulation.

→ ① Optimization via simulation refers to the problem of maximizing or minimizing the expected performance of a discrete event, stochastic system that is represented by a computer simulation model.

② optimization usually deals with problems with certainty, but in stochastic system that is represented by a computer simulation model.

③ let x_1, x_2, \dots, x_m be the m controllable design variables and

$y(x_1, x_2, \dots, x_m)$ be the observed simulation output performance on one run.

④ To optimize or ~~at~~ $y(x_1, x_2, \dots, x_m)$ with respect to x_1, x_2, \dots, x_m is to maximize or minimize the mathematical expectation of performance.