Computer Science and Applied Mathematics

### ANALYSIS AND SYNTHESIS OF COMPUTER SYSTEMS

Gelenbe/Mitrani

# Analysis and Synthesis of Computer Systems

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## Preface

The subject of computer performance evaluation has grown steadily in the last decade or so. As the area of computer applications expanded out of the scientific domain and into those of business, commerce and administration, it became more and more important to know in advance what a given system can do and how it should be operated. After all, an inefficient computer system can cause no small amount of chaos—from pay-slips not arriving on time, to traffic jams in a busy town centre. There are also cost considerations: large computer systems represent considerable investments, both for the manufacturers who design and build them, and for the users who buy them. It is in the interest of everyone concerned to be able to answer questions about the performance of such systems—preferably before they are built or bought. This is our interest, too.

The approach we adopt is that of probabilistic modelling; we deal with the construction and analysis of models, with their use in evaluating the performance of computer systems and with their application to system design and control. Why use probabilistic modelling in relation to objects which are man-made and deterministic by nature? The answer is, of course, that while the machines are deterministic, the demands placed upon them are not. The arrival patterns and the resource requirements of the tasks comprising the demand are, in general, random. Each task execution involves such a complex sequence of operations (machine instructions) that it is, for all practical purposes, unpredictable. We model these random phenomena by introducing probabilistic assumptions.

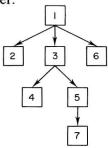
The problems that we consider fall into two broad categories: problems of analysis and problems of synthesis. If all system parameters—number and speed of processors, size of memory, characteristics of demand, system software, etc.—are given, and the question is "how does this

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system perform?", then we have a problem of analysis. If, on the other hand, some components of the system are regarded as variable (typically these are concerned with job scheduling and resource allocation algorithms), and one is asked to find a configuration that achieves a certain pre-specified performance objective, then we say that the problem is one of synthesis. Often, although not always, analysis and synthesis go hand in hand. Suppose, for example, that we are offered a family of job scheduling strategies and are asked to choose the best one according to some criterion (a synthesis problem). In some cases, the answer can be obtained directly from general principles; in others, it is necessary first to analyse the strategies in the family and to evaluate their performance. We shall encounter both situations.

The reader is introduced to the methodology and the necessary tools of probabilistic modelling in Chapter 1. Some basic results from the theory of Markov processes and queueing theory are derived there. In Chapter 2, a class of queueing systems known as "systems with a server of walking type" are analysed and the results are applied to the performance evaluation of several peripheral storage devices. Chapter 3 is devoted to the analysis of queueing network models. Cases with known closed-form solutions are examined, involving fixed and variable job populations, and allowing multiple job classes. Chapters 4 and 5 are concerned with the approximate analysis of models for which exact solutions are not available. Chapter 4 describes diffusion approximation techniques; these are applied to both single-node and network models. Chapter 5 deals with decomposition and iterative methods aimed at reducing the complexity of models. Chapters 6 and 7 address synthesis problems in single and multiple resource systems: conservation laws, achievability of response time objectives, minimisation of certain cost functions, control of the degree of multiprogramming in virtual memory systems with one or several job classes and control of performance by memory allocation.

The box diagram illustrates the dependencies among chapters. An arrow from one box to another indicates that the corresponding chapters should be read in that order.



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Chapters 2 and 4 have a somewhat higher mathematical content than the others.

The book is aimed at students, researchers and practitioners in the field of performance evaluation. However, it is not intended as a "book of recipes": we hope, rather, that sufficient knowledge and understanding are imparted to the reader so that he can write his own recipes. Knowledge of basic computer system concepts is assumed, as well as some mathematical background, including undergraduate calculus and probability theory.

January 1980

E.G.

I.M.

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# 1 Basic Tools of Probabilistic Modelling

#### 1.1 General background

On a certain level of abstraction, computer systems belong to the same family as, for example, job-shops, supermarkets, hairdressing salons and airport terminals; all these are sometimes described as "mass service systems" and more often as "queueing systems". Customers (or tasks, or jobs, or machine parts) arrive according to some random pattern; they require a variety of services (execution of arithmetic and logical operations, transfer of information, seat reservations) of random durations. Services are provided by one or more servers, perhaps at different speeds. The order of service is determined by a set of rules which constitutes the "scheduling strategy", or "service discipline".

The mathematical analysis of such systems is the subject of queueing theory. Since A. K. Erlang's studies of telephone switching systems, in 1917–1918, that theory has progressed considerably; today it boasts an impressive collection of results, methods and techniques. Interest in queueing theory has always been stimulated by problems with practical applications. In particular, most of the theoretical advances of the last decade are directly attributable to developments in the area of computer systems performance evaluation.

Because customer interarrival times and the demands placed on the various servers are random, the state S(t) of a queueing system at time t of its operation is a random variable. The set of these random variables  $\{S(t), t \ge 0\}$  is a stochastic process. A particular realisation of the random variables—that is, a particular realisation of all arrival events, service

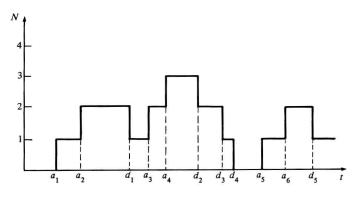


Fig. 1.1

demands, etc.—is a "sample path" of the stochastic process. For example, in a single-server queueing system where all customers are of the same type, one might be interested in the stochastic process  $\{N(t), t \ge 0\}$ , where N(t) is the number of customers waiting and/or being served at time t. A portion of a possible sample path for this process is shown in Fig. 1.1: customers arrive at moments  $a_1, a_2, \ldots$  and depart at moments  $d_1, d_2, \ldots$ 

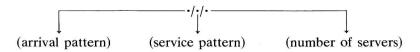
An examination of the sample paths of a queueing process can disclose some general relations between different quantities associated with a given path. For instance, in the single-server system, if  $N(t_1) = N(t_2)$  for some  $t_1 < t_2$ , and there are k arrivals in the interval  $(t_1, t_2)$ , then there are k departures in that interval. Since a sample path represents a system in operation, relations of the above type are sometimes called "operational laws" or "operational identities" (Buzen [1]). We shall derive some operational identities in section 1.7. Because they apply to individual sample paths, these identities are independent of any probabilistic assumptions governing the underlying stochastic process. Thus, the operational approach to performance evaluation is free from the necessity to make such assumptions. It is, however, tied to specific sample paths and hence to specific runs of an existing system where measurements can be taken.

The probabilistic approach involves studying the stochastic process which represents the system. The results of such a study necessarily depend on the probabilistic assumptions governing the process. These results are themselves probabilistic in nature and concern the population of all possible sample paths. They are not associated with a particular run of an existing system, or with any existing system at all. It is often desirable to evaluate not only the expected performance of a system, but

also the likely deviations from that expected performance. Dealing with probability distributions makes this possible, at least in principle.

We shall be concerned mainly with steady-state system behaviour—that is, with the characteristics of a process which has been running for a long time and has settled down into a "statistical equilibrium regime". Long-run performance measures are important because they are stable; being independent of the early history of the process, and independent of time, they are also much easier to deal with. We shall, of course, be interested in the conditions which ensure the existence of steady-state.

This chapter introduces the reader to the rudiments of stochastic processes and queueing theory. Results used later in the book will be derived here, with the emphasis on explaining important methods and ideas rather than on rigorous proofs. In discussing queueing systems, we shall use the classic descriptive notation devised by D. G. Kendall:



e.g. D/M/2 describes a queueing system with Deterministic (constant) interarrival times, Markov (exponential) service times and 2 servers.

#### 1.2 Markov processes. The exponential distribution

Let S(t) be a random variable depending on a continuous parameter  $t \ (t \ge 0)$  and taking values in the set of non-negative integers  $\{0, 1, 2, \ldots\}$ . We think of t as time and of S(t) as the system state at time t. The requirement that the states should be represented as positive integers is not important; it is essential that they should be denumberable. Later, we shall have occasions to use vectors of integers as state descriptors.

The collection of random variables  $\{S(t), t \ge 0\}$  is a stochastic process. That collection is said to be a "Markov process" if the probability distribution of the state at time t+y depends only on the state at time t and not on the process history prior to t:

$$P(S(t+y)=j \mid S(u); u \le t) = P(S(t+y)=j \mid S(t)), \quad t, y \ge 0, j=0, 1, \dots$$
(1.1)

The right-hand side of (1.1) may depend on t, y, j and the value of S(t). If, in addition, it is independent of t, i.e. if

$$P(S(t+y) = j \mid S(t) = i) = p_{i,j}(y) \text{ for all } t,$$
 (1.2)