## THE EFFECT OF ORDER SIZE ON THE OPERATION OF A HYPOTHETICAL JOB SHOP MANUFACTURING SYSTEM

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SYSTEM

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## PREFACE

This investigation was based upon the idea that the operation of a job shop manufacturing system is affected by the sizes of orders processed through it. The approach was to build a hypothetical job shop with well defined capabilities and to test its reaction to different order sizes and different mixtures of order sizes. Criteria were established to detect any differences in the reactions of the system to the various test conditions.

The literature search failed to reveal any instance where the relationships between order sizes and job shop system performance were treated explicitly. The usual approach was to account for order size by postulating distributions of machine center flow times and sampling from these distributions for each order. Order size, then, was implicitly included in the amount of time required to process an order by a center. By contrast this investigation generates machine center flow times as a function of order size。

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## CHAPTER I

## INTRODUCTION

The purpose of this dissertation is to describe the results of investigating the reactions of a hypothetical job shop manufacturing system to controlled variations in the attribute of size of orders passing through the system. The investigation concentrates attention on four measurable reactions of the system to changes in order size; idleness, order flow time, delivery time, and waiting time。

A computerized model of the system is developed and twenty-five simulations performed to generate observations under five mixes of order sizes and five conditions of setup time. Seven corollary simulations are run to test the validity of the assumption of certain equilibrium conditions in the model.

The results of this research indicate that increases in the sizes of orders processed by the job shop manufacturing system:

1. increases production in the job shop by reducing the incident of setup,
2. increases the total flow time of orders through the system in proportion to the increase in job size,
3. enlarges the means and variances of all time related distributions in the system,
4. does not materially alter the shapes of the time related distributions.

The contributions of the research are considered to be four in number. First, an estimating technique is devised to predetermine the mean time between input of jobs to the system. The technique appears to eliminate the need for service rate runs. It is probably best suited to simple systems such as the one investigated. If this is true, it has limited application。

Second, the technique of permitting one element of center flow time, queue time, to be generated as a function of the operation of the system appears to be a sound approach not noted in the literature. The technique, when refined, should permit the derivation of estimators for center flow time in systems whose records are confined to setup and processing times.

The analysis of idle time into two components reveals an opportunity to reduce idle time in the system by causing the two components to coalesce. Segmenting idle time into components of idleness caused by absence of work and idleness caused by setup makes clear the potential reduction in idle time by the expedient of a procedural change in the management of information in the job shop manufacturing system.

Finally, the investigation tends to confirm the researcher's understanding of current theory while perhaps adding a small increment of knowledge to it. What seems to be worthwhile is not that confirmation takes place, but that it is achieved by employing what is considered to be a refined technique in modeling machine center flow time in job shop systems.

The remainder of this chapter is devoted to antecedents; defining the system and reporting the results of the literature search of previous, allied investigations. Chapter II describes the computerized model constructed for the research effort. Chapter III describes the experiment in detail, formalizes the hypotheses tested, and displays the rationale for the various choices required of the researcher. Chapter IV presents the discussion and analysis of the outcomes of the experiment and the inferences drawn. Finally, Chapter $V$ summarizes the results and conclusions, suggests future research, and discusses the reservations about the results of this research effort.

## The Job Shop Manufacturing System

The job shop manufacturing system is distinguished by several usually well understood characteristics. The purpose of these next several sections is to describe these characteristics, the variety of ways they might be viewed, how they contrast with the characteristics of other manufacturing systems, the extent to which they have been
treated in the past, and how they are viewed in this research effort.

Job Shop Defined

A job shop is a manufacturing system composed of differentiated work centers (23). This means that processing capability is homogeneous within centers and heterogeneous among centers. As several authors describe it, machines are grouped together in centers according to like function (27, 34). It has been observed, however, that machine centers have evolved where the function of two centers is the same, but the means of control of the machines is different, e.g., one center in a shop is composed of numerically controlled milling machines and another is composed of manually controlled milling machines.

Job shop systems are more likely to have general purpose rather than special purpose machines. It is not necessary, however, that this characteristic be inviolate. It should be expected that the smaller the system or the more diverse the demands on the system the more likely that all machines have a wide range of capabilities. Even this comment is subject to interpretation. For clarity, consider the activity of milling. Generally, this is thought of as shaping or dressing metal by passing the metal by revolving cutters of various sizes or shapes. If a machine can make only one cut or if it can make several cuts but only with one degree of freedom, it is a special purpose
machine. If it is more versatile and can be set to make a variety of cuts with several degrees of freedom, it is a general purpose machine. A more exact distinction seems unnecessary.

The pure job shop system is characterized by manufacturing on demand to customer order. Its activity is not buffered or protected from fluctuations in demand as is the case with other manufacturing systems. Perhaps the best contrast of the job shop system from this point of view is with the repetitive manufacturing system。 Here there exist a fairly well defined range of products and the means of forecasting future demand for these products. Machines may still be grouped by function although their relative physical location is probably influenced more by an established technological order of processing activities required for the products than is the case of the job shop. The prior knowledge of most of the products of the system and expected demand for the products provides the opportunity to manufacture to inventory rather than exclusively to customer order. This may be accomplished in one or both of two ways.

Products are assumed to be composed of component parts. These component parts may be manufactured according to some repetitive schedule and held in inventory pending customer order. When the order arrives, the parts are assembled and the product shipped This option is common when products differ in final configuration but are basically the same. A good example is an accounting machine such as those
produced by The National Cash Register Co., Dayton, Ohio. The other option is to manufacture to finished inventory. This procedure also employs a repetitive manufacturing schedule, but it differs in that products are completed and stored to meet demand.

There are, of course, a variety of ways in which these two options may be combined. Of major interest is the point that the repetitive system employs inventory to decouple demand from supply; hence, tends to provide for less fluctuation in the manufacturing activity. By contrast, the job shop system does not manufacture to inventory; hence, its activity is directly related to demand and may be highly volatile.

Another distinguishing characteristic of the job shop system is its general inability to cope systematically with the scheduling of jobs through the system (14). In the repetitive manufacturing system it is often possible and profitable to identify a production cycle in. which the machine sequence, job sequence, and product run length are specified and are repeated. There is no such neat array of tasks in the job shop system. As a consequence, scheduling is almost a continuous process. When the first center in the ordered set of centers selected to process a job is free, it is the usual practice to release the job immediately to the system. If the center is not free, the job may enter the queue at that center or be diverted to an alternate processing route when technologically feasible.

There are several problems associated with random-job scheduling. As a result optimal policies for scheduling are difficult to formulate and even more difficult to defend.

In partial summary, the job shop manufacturing system is composed of differentiated work centers. The system operates only on demand and then to customer specification. It does not have any prior knowledge of when, what, or in what amount it is expected to produce, except within the scope is its advertised capability. (An industrial grinding company, for example, would not expect to be asked to extrude metal.) There are many technologically feasible routes through the system. These are both a function of the nature of the job and the existence of technically correct alternative ways of doing it. As a result, jobs interfere with one another (compete for machine time) and delays occur.
Balancing

The balancing problem deals with the equality of output of each successive operation in the sequence of a line (5). Its job shop counterpart is relative equality of output of machine centers. In both cases the desired solution to the problem means reduced interruption of work at downstream stations and elimination of excessive backlogs at any one station.

Practical solutions to the problem of maintaining balance in the job shop include the selective use of over-
time, installing more capable or simply more machines, rerouting orders around centers with large backlogs, and alteration of machine loading. Machine loading, the amount of work to be accomplished by a machine, usually measured in time units, has been most often treated for the production line with continuous or repetitive manufacturing activities or a job shop with repetitive production. An early work by Salveson (31) employed linear programming to find optimal loading in what he calls a quasi-job-shop. The assembly line balancing problem has been treated by several authors (15, 17, 28, 35)。

## Routing

Routing determines where work is to be performed. Routing is also called technological routing, technical requirements, etc. Implicit in this description of routing is a requirement to consider the order as well as the nature of work for any given job. For example, cutting must be accomplished before polishing. Other like kinds of technical order requirements exist. The various models examined in preparation for this research did not deal with the routing problem. Rather it was assumed that routing was predetermined and fixed outside the job shop system. The other alternative, of course, is to postulate and employ alternative technologically correct routings for each job and to establish internal system rules for choosing among the alternatives.

## Scheduling

Scheduling determines when work is to be accomplished. Usually scheduling is used as an inclusive term meant to describe a rather precise and complete planning function Several jobs and several machines are considered simultaneously. Machine loading, routing, sequencing (to be discussed), materials, labor, etc. are jointly considered and jobs and machines are mjxed in some best way. Usual criteria deal with the concept of efficiency; e.g., maximum use of available production time.

Scheduling as just described is not particularly appropriate to the job shop system. As observed on page 6, job shop scheduling is an almost continuous process. Additionally, it covers the whole spectrum of tasks starting with drawing materials and ending with completion of the customer order.

## Sequencing

The sequencing problem, sometimes called the schedulesequence problem, deals with the question of when to produce an order, not with respect to the clock, but with respect to other orders. The problem has been solved for continuous manufacturing systems, but not for job-shop systems (34).

Sequencing in the job shop usually has been approached by periodically adjusting the relative order of jobs waiting to be processed in the various queues in the system. Rules for making such adjustments and the criteria for choosing
among them are all concerned with some function of the time a job stays in the system. Perhaps the most exhaustive research to date on sequencing rules for an idealized job shop is that reported by Conway (9). He compares and evaluates 17 basic rules plus 23 variations and combinations of these basic rules, all with different values of the control parameters. In total, he tested 92 different rules.

As a note of possible interest he did not test the rule employed in this study which is described in Chapter II. He did, however, test a modified version of SLACK (slack time rule) which is conceptually similar. The SLACK rule gives preference to the job with the least time remaining until the due date after deducting the remaining processing time. SLACK is defined as follows:

$$
P_{i}=D_{i}-T-\sum_{j=k}^{M_{i}} P_{i j}
$$

where
$P_{i}$ priority at the ith job
$D_{i}$ due date of the ith job
$T$ time at which a selection of machine assignment is made
$P_{i j}$ processing time required at the $j$ th center for the ith job
$j$ index over the sequence of machine centers
k the next center
$M_{i}$ the total number of centers for the ith job

Conway's modification of the rule involves weighting the resulting $P_{i}$ by dividing it by the number of remaining centers and giving priority to the job with the smallest ratio of slack/center remaining.

Setup time (time to prepare a machine to process a job) and the sequence of jobs processed by the machine may be related. Consider two jobs $A$ and $B$. If the sequence $A B$ results in setup time $S_{A B}$ and the sequence $B A$ results in setup time $S_{B A}$ and $S_{A B}<S_{B A}$, the sequence $A B$ is preferred. This is equivalent to stating that setup time is a function of the machine, the job, and its relationship to other jobs in the stream passing through the machine. No meaningful examination of this job dependent characteristic of setup time was discovered although several authors indicate an awareness of it.
Dispatching

Dispatching is determining the time an order is released to the job shop system so that work on it may begin. It is, in effect, a decision to permit the order to compete for machine time with orders already in the system. Some authors define dispatching to include issuing instructions about the order as it proceeds through the system. However, this function is thought to be well covered under the sequencing concept.

## CHAPTER II

DESCRIPTION OF THE MODEL

The model represents a job shop manufacturing system with a small number ( 1 to 20 ) of single machine centers. Each center may be made different from or identical to any other center. Each can process one and only one job at any one time. One job may not preempt another. Jobs consist of units of product which are identical both within and among jobs. The time required to process a job is a function of its magnitude in units and the center assigned to process it. The time required to prepare a center to process a job (setup time) is a function of the center. The time a job waits to be processed at any center is a function of the number and magnitude of higher priority jobs also waiting or being processed.

## Machine Center Logic

Each center can process one and only one job at any one time. The time required to process a job through a center is the product of the magnitude of the job in units and the unit processing time. The unit processing time is a specified random variable.

Each job requires that the center assigned to process it be setup. This implies that the job is always different from the job immediately preceding it through the center. Setup time is a function of the center and is a specified random variable.

The time a job waits to be processed depends upon the number of higher priority jobs also waiting or being processed. Hence, queue time is a generated random variable dependent upon the utilization of the center.

The time to process a job by a center, T, is the sum of three random variables; the time the job waits, $Q$, the time required to prepare the center to process the job, $S$, and the time required to process it, $P$. The time required to move a job from one center to the next center is considered to be included in the waiting time at the next center. Waiting time may be zero. Setup time may be specified as zero to simulate operations for which no setup is necessary. Unit process time is always greater than zero and never less than one clock unit per unit of product. However, the sampling technique employed provides for effective unit process time of less than one clock per unit of product. For example, suppose the job is of size loo units and that the job is to be processed by a center with a unit process time of one. The product of 100 units and one clock unit per unit equals 100 clock units. This length of time is taken as the mean of the population of process times. Suppose further that a sample of size one from this
population produces a process time of 95 . This results in an effective unit process time of 0.95 clock units.

Center operation probably can best be described with the aid of the schematic in Figure 1. Recall that the time required to process a job through a center is $T=Q+S+P$ 。 Q, S and P, as listed, also provide the order of events within the center. The center is represented by the large block T. A job to be processed enters the center through block $Q$. It goes directly to the "on-deck" block $Q_{1}$. If $Q_{1}$ is empty and block $P$ is idle, it moves to block $S$, then to block $P$ and exits the center, $T$.


Figure 1. Machine Center Flow, Center T

If $Q$ is empty and $P$ is operating, the $j o b$ is held in $Q_{1}$ until $P$ is idle $\quad$ If $Q_{j}$ is occupied, the priority of the incoming job is compared with that of the job occupying $Q_{1}$. The lower priority job is sent to $Q_{2}$. Blocks $Q_{1}$, $S$ and $P$ can hold only one job at a time; $Q_{2}$ is unrestricted。

To summarize briefly, each machine center is composed of four blocks. Two of these, $Q_{1}$ and $Q_{2}$, simulate the waiting line; block $S$ simulates the machine setup activity; and block $P$ simulates the processing activity. The status of block $P$ controls the access to block $S$. The status of block $Q_{1}$ and the priority of the job in $Q_{1}$, if any, determine whether a job proceeds to block $S$ or to block $Q_{2}$.

## Queue Discipline

The model employs a job sequencing algorithm developed by Fabrycky and Shamblin (20). The algorithm provides a way to change the sequence of jobs waiting in the various queues in the system according to their relative urgency. This is accomplished periodically, for each order, by a standardized comparison of the due date of the order, the current date, and the expected processing time of the remaining machine centers assigned to the order.

The algorithm is based upon properties of the Central Limit Theorem. If $\mu_{j}$ and $\sigma_{j}^{2}$ are the mean and variance of order flow times through the jth center, the total flow time of the ith order through the shop, $T_{i}$, is approximately normally distributed with mean

$$
\mu_{i}=\sum_{j=1}^{n} \mu_{j}
$$

and variance

$$
\sigma_{i}^{2}=\sum_{j=1}^{n} \sigma_{j}^{2} .
$$

The more centers assigned to process jobs, the more nearly the distribution of $T_{i}$ corresponds to the normal distribution。

Suppose the ith order is at machine center $k, k=1$, 2, ---, $n$. The mean flow time before completion of the order is

$$
\sum_{j=k}^{n} \mu_{j}
$$

The flow time variance is

$$
\sum_{j=k}^{n} \sigma_{j}^{2} .
$$

The expression

$$
z_{i}=\frac{\left(D_{i}-C\right)-\sum_{j=k}^{n} \mu_{j}}{\sqrt{\sum_{j=k}^{n} \sigma_{j}^{2}}}
$$

is the standardized value of the distribution of remaining flow time where $D_{j}$ is the due date of the ith order and $C$ is the current date. The values of $z$ determine the positions of their respective orders in the machine center queues. Implicit in these $z$ values are the probabilities of meeting the due dates. The order with the algebraically smallest $z$ implies the smallest probability of completing the order by its due date. Hence, this order will be positioned in a queue ahead of orders whose $z$ values are larger.

The effect of this queue discipline rule is similar to the effect of an expediter who employs current knowledge of the state of the system and jobs in progress to decide the order of near term processing activities. The rule tends to equalize the probabilities of all jobs being completed by their due dates. Implicit in the employment of this rule is the assumption that the value of completing a job on time is the same as the value of completing any other job on time.

## System Service Rate

The service rate of the system is defined to be job output rate when all machine centers in the system are operating at maximum possible capacity. One hundred percent utilization of the processing capacity of the system is possible only when no setup time is required at any center.

For an unstructured system; i.e., a system in which the routes for orders are selected at random by sampling from a uniform distribution, it is possible to estimate the system service rate.

In this model, the number of machine centers in any route are equally likely. If the system contains ten centers, the probability that a route contains one center is the same as the probability that it contains 2, 3 or 10. In other words, the probability of the number of centers in a route for any order for a system with ten centers is 0.7. The expected number of centers in a route from this system is

$$
E[n]=\begin{gathered}
10 \\
\sum n \\
\frac{n=1}{10}
\end{gathered}=5.5
$$

It is not true that routes through the system are equally likely. The method of choosing the number of centers in each route precludes this. There are ten ways to have routes containing one center and $10^{10}$ ways to have routes containing ten centers. Since returns are permitted, in general, there are $\sum_{n=1} 10^{n}$ (more than 1.1 billion) routes through the system. If it were true that the routes were equally likely, the expected number of centers in ahy route would be in excess of 9.9.

Consider the system with zero setup times. A production day is defined as 1000 clock units. The real time equivalent is approximately 28.8 seconds per clock unit. In a system of ten machine centers there are a maximum of 10,000 production clock units available per day. Suppose the system processes jobs of size 100 units and that the unit processing time is one clock unit at all centers. It is easy to see that product of the expected number of centers and the expected processing time per job per center will result in the expected time per job through the system since these are independent events. Hence, for this case the expected flow time through the system is

$$
5.5(100)=550 \text { clock units/job. }
$$

Since there are 10,000 clock units available, the expected service rate must be $10000 / 550$, or about 18 jobs per day. In this simple case, then, for jobs of size m with unit process time of $t$, processed by a system of size $n$, the expected service rate, $u$, can be estimated as follows

$$
\hat{\mu}=\frac{1000 n}{\frac{m t}{n} \sum_{j=1}^{n j}}
$$

or

$$
\hat{\mu}=\frac{1000 n}{m t E[n]}
$$

where $E[n]$ is the expected number of centers per order. When setup time is greater than zero and equal at all n centers, $\mu$ can be estimated as follows;

$$
\hat{\mu}=\frac{1000 n}{(m t+s) E[n]} .
$$

Using the previous values and setting s = 10 clock units per order per center,

$$
\hat{\mu}=\frac{1000 n}{[100(1)+10] 5.5}=16.5 \text { orders/day. }
$$

A third case arises when orders are of different, but known sizes. Suppose two sizes of orders are processed by the system and that the percentage of time each order occurs is known. If half the orders are of size $m_{1}$ and half are of size $m_{2}$, the service rate calculation is

$$
\hat{\mu}=\frac{1000 n}{\left.\frac{\left[m_{1} t+m_{2} t\right)}{2}+s\right] E[n]}
$$

Again, using the previous values and setting $m_{1}=100$ and $m_{2}=10$

$$
\begin{aligned}
\hat{\mu} & =\frac{1000(10)}{\left[\frac{100(1)+10(1)}{2}+10\right]} 5.5 \\
& =27.9 \text { orders/day. }
\end{aligned}
$$

In general, then

$$
\hat{\mu}=\frac{1000 n}{(E[m t]+s) E[n]} .
$$

Finally, when setup time is allowed to vary among centers, the computation becomes

$$
\hat{\mu}=\frac{1000 n}{(E[\mathrm{mt}]+E[s]) E[n]}
$$

## Establishment of Job Due Dates

Due dates are a function of job size, technical processing requirements, system performance, and management's interest in on-time deliveries.

Job size, the number of units of product in an order, partially determines the system flow time distribution from
which future system performance is estimated. Technical processing requirements, the number and sequence of centers needed to process the job, are accounted for by assuming that all permutations of machine centers are feasible. Management's interest in on-time deliveries is reflected explicitly by considering the variation in system performance.

Due dates are established in accordance with

$$
D_{i}=R_{i}+\sum_{j=1}^{n} \mu_{j}^{\prime}+z_{i} \sqrt{\sum_{j=1}^{n} \sigma_{j}^{\prime 2}}
$$

where $R_{i}$ is the release date of the ith order. In the model, jobs are released as soon as they arrive.. The passage of time to contract for the order, prepare specifications, coordinate delivery of materials, etc. is assumed to have occurred previously. If the total flow time of the ith order is approximately normally distributed with mean

$$
\mu_{i}^{\prime}=\sum_{j=1}^{n} \mu_{j}^{\prime}
$$

and variance

$$
\sigma_{i}^{\prime 2}=\sum_{j=1}^{n} \sigma_{j}^{\prime 2}
$$

then

$$
D_{i}=R_{i}+\mu_{i}^{1}+z_{i} \sigma_{i}^{1}
$$

where the prime designates parameters of populations of
times by order size. Now $z_{i}$ may be chosen so that management is satisfied with the probability of on-time delivery (29) .

The model permits five choices of $z_{i}$ and the means of selecting them according to any distribution. This capability is useful to the extent that it provides a means of simulating underestimating and overestimating system performance, promising due dates which cannot be met, or other deviations from policy.

Job Sizes

Five job sizes are possible. One of these, NTYPE(3) energizes a TRACE block. Consequently, it is possible to record the complete history of all NTYPE(3) jobs as they proceed through the job shop. This feature is useful as a diagnostic tool in the early stages of manipulating and testing the model. An example of the TRACE report is contained in Appendix A.

The main reason for providing for various job size inputs is to test the effect of different job sizes on the operation of the job shop manufacturing system; the purpose of this research.

Job size mixes may be chosen in any proportion desired. Job sizes may be any integer value greater than or equal to one and less than $2^{15}$.

## Demand on the System

Demand on the job shop manufacturing system may be created by drawing from a distribution of demand with job sizes subsequently assigned by sampling from a distribution of jobsizes.

The mean arrival rate must be less than or equal to the system service rate to prevent the building of infinite queues. Since it is possible to estimate the service rate of the system with reasonable accuracy, service rate runs don't appear to be absolutely necessary.

## Periodic Status Reports

Periodic status report capability has been built into the model to provide for examination of the state of the system at intermediate points during a simulation. A status report is available as often as once at the end of each day or it may be suppressed entirely during a simulation. An example of the status report is presented in Appendix B.

The primary value of this feature of the model is in providing a way to observe the rate at which the model achiéves steady state, the functioning of the random number generators, the growth of some of the various statistics recorded at the end of a simulation, and a way of comparing reactions according to other than terminal run conditions. During diagnostic runs, it provides an additional means of pinpointing error sources.

## Statistics

In addition to the information available from the model through the TRACE report and the STATUS reports, the model generates a variety of statistical tables. Some of these are provided automatically by the General Purpose Systems Simulator II. Others are unique to this model.

The output consists of 53 tables:
Tables Tabulated by Frequency Class

1-10
21-30
41-50
61-65
$66-70$
71-75
$76-80$
81
82
83

Center Flow Time
Center Idle Time
Center Queue Time
System.Flow Time by Size Type
D-A Time by Size Type
System Flow Time by z Type
D-A Time by z Type
System Flow Time
System D-A Time
System Inter-exit Time

Examples of these tables are contained in Appendix C.
Each table contains the distribution of the observed frequency of occurrence of values of a system variable or function of a system variable. These are recorded by frequency class. There is no limit on the number, incremental size, or range of frequency classes except that resulting from computer space allocation. In addition to the frequency distribution, (which may be in the form of
weighted entries) each table provides the total number of entries in the table, the mean, and the standard deviation.

## Variables and Rules

The description of the model thus far indicates that it is possible to control two variables. These are the values of the initial z and the sizes of orders. Choosing a positive value of the initial z corresponds to a management decision to contract for due dates which will enhance the probabilities of completing orders on time. Choosing the sizes of orders to be processed by the system implies both the capability and the reason for combining or splitting orders to improve system performance. For this research, the only decision variable is taken to be the choice of the order or job sizes. The choice of initial z with minor perturbations is employed as an unchanging rule by which orders are released to the system.

To recapitulate, the variable under the control of the decision maker is the size of the order in homogeneous units of product. All other variables either are assigned magnitudes based upon what may be regarded as preestablished rules for repetitive decision situations, or they are considered to be variables describing the nature of the environment and the system and outside the control of the decision maker.

Events occur in chronological order. Orders arrive according to some distribution of demand. They are assigned
the number of centers to process the order; then they are assigned to specific centers, both actions by sampling from the uniform distribution. Each order is given a due date and it is released to the system. During its processing it competes for machine time at each center according to the value of its urgency number, $z$. When it has been processed by all assigned centers, it departs the system and appropriate statistics are recorded. This process is repeated for all orders until the simulation is terminated. Termination may be accomplished in one of two ways; time, or orders processed. In the experiment reported in this paper, termination is accomplished by controling run lengths (time).

## CHAPTER III

## THE EXPERIMENT

The method of experimentation with the model is to make changes in selected variables and then to analyze the effects of these changes upon the behavior of the job shop manufacturing system. In order to study the results in some systematic way, it is necessary to decide upon the proper method or strategy for analysis. Such considerations are the subject of this chapter.

The Delimited System

Chapter II describes a computerized model with the capability of simulating any number of job shop manufacturing systems with similar characteristics. It is now necessary to define one or more with which to experiment. This is accomplished by making a number of choices. These include the number of centers to be in the system, the operating characteristic of each center, and the period of the queue discipline rule. The important effect of the second of the three choices is the decision to employ a number of identical or different machine centers in the system. It appears to be the most critical of the choices and will be discussed at some length. The other two choices can be
dispatched quickly and will be treated first.
Ten machine centers are to be employed in the system. The selection of this number of centers is not entirely arbitrary. There are four practical, if not important, reasons for selecting ten. First, it is a convenient factor, thus facilitating computational effort. Second, the queue discipline rule was first tested in a system of ten centers. Curiosity dictates the same size sys.tem to see if comparable results obtain. Third, diagnostic runs with the model proved the computer to be extremely slow, thus placing a high cost in computations per center in the system. Finally, ten centers appear to be a sufficient number to create the kind of interference and competition for machine center processing time thought to be present in real systems.

The period for the queue discipline rule; i.e., the period of time permitted to elapse before the urgency numbers are recomputed for each of the orders in the system, is taken as one day. The urgency numbers are computed and the orders realigned in the ten queues in the system at the end of the work day and before the beginning of the next work day. It would be possible, of course, to choose other intervals of time between updating the positions of orders in the queues, but there seems to be no compelling reason to do so.

The question of the operating characteristics of the individual machine centers in the system appears to be of
substantially more importance than the other choices just discussed. First, should the centers be identical or different and why? Second, should machine center processing time per unit of product and setup time per order be taken as constants or random variables? And third, if they are taken as random variables, what function or functions should be employed to assign value to each sample point?

Building a model of a hypothetical system doesn't appear to sever the researcher from all connection with reality. At a minimum, the hypothetical system ought to be a reasonable representation of a possible real system. While no claim of general applicability of the results of this research will be made, the possibility of such application should not be foregone for lack of reasonableness. In this same vein, the delimiting choices are thought to result in a suitable system for study. This kind of belief cannot, of course, be completely validated. What can and will be done is to display the choices and the rationale for them for separate examination:

In addition to the stated need for reasonableness is a need for simplicity, at least to the extent that the opposite, complexity, may tend to camouflage sought after answers. Simplicity is not necessarily achieved at the expense of reasonableness or validity. All models are simplifications to some degree and this one is not an exception. Neither complexity nor simplicity are necessary conditions for validity. The acid test of the validity of a
model is its ability to predict so the degree of complexity of the model is only important to this end, if.at all. It appears, then, that simplicity is not antithetical to validity but preferred for the different reason of visibility. In other words, simplicity is desirable to increase the probability of seeing answers; reasonableness is desirable to increase the probability of the applicability of those answers. These two points of view are intended as general arguments in support of the remaining choices.

The system is taken as a set of ten identical machine centers analogous to a network of identical single-server queues. This system, and the arguments for it, are much like that employed in the previously cited work of Conway (9). One notable difference is the inability to postulate distributions of service times until after the fact of simulation since service times (center flow times) are generated as the sum of three random variables only two of which are specified. The primary benefit accruing through the use of identical centers, at least with respect to the attribute of time, is a symmetrical or balanced system. This balanced condition eliminates the need to introduce ways to combat inbalance leading to excessively large individual queues or excessive idle time at downstream centers. Additionally, starting with a balanced system portends no loss of generality since inbalance would have to be corrected in any event.

The remaining choices are discussed jointly. As will
be recalled, these deal with the matter of variability in the operations at each center - the nature of the distributions of setup and processing time. It was decided that both should be treated as random variables rather than as constants, if for no other reason than to be consistent in acknowledging the stochastic nature of real systems. This choice is not judged critical since the sum of a constant and a random variable remains a random variable Hence, one statistic of interest, center flow time ( $T=Q+S+P$ ) will be a random variable regardless of which choice is made. Finally, setup time, $S$, is specified as a uniformly distributed random variable

$$
\begin{aligned}
f(S) & =\frac{1}{b-a} \quad a<S<b \\
& =0 \quad \text { otherwise }
\end{aligned}
$$

with parameters $a=.9 S, b=1.1 S$ and $E[S]=(0,50,100$, 250, 500). Process time, P, is specified as a uniformly distributed random variable with the same treatment of the parameters a and b and with $E[P]=100$ or $E[P]=500$ corres ponding to the size of the order being processed.

Since center flow time has been identified as a statistic of interest, and since the choice of distributions from which to draw setup and process time may appear questionable, the results will be displayed and argued here

Figure 2 is the continuous analogue of a typical discrete distribution of center flow time, $T$, from one of the 32


Figure 2. Typical Distribution of Center Flow Time (T)
experimental runs which serve as the information base for this research. Note that the center flow time distributions generated in these simulations have a form very similar to those discovered in research with a currently operating job shop manufacturing system. This information is an unpublished observation, according to the author, of the research reported on the development of the probability based sequencing algorithm (20).

## Demand on the System

Jobs are released to the system one at a time in order and when generated. Interarrival times are obtained by sampling from an exponential distribution with the mean set to yield a nominal system utilization of 90 percent (hence a utilization of 90 percent at each center) under each of the 25 conditions chosen for the experiment. These interarrival times are displayed in Table I. They were pre-

TABLE I
MEANS OF EXPONENTIAL INTERARRIVAL TIMES
TO PRODUCE 90 PERCENT UTILIZATION

| Mean <br> Setup <br> Time | $(1) 100$ <br> $(0) 500$ | $(.75) 100$ <br> $(.25) 500$ | $(.5) 100$ <br> $(.5) 500$ | $(.25) 100$ <br> $(.75) 500$ | $(0) 100$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 60 | 120 | 180 | 240 | 300 |
| 50 | 90 | 150 | 210 | 270 | 330 |
| 100 | 120 | 180 | 240 | 300 | 360 |
| 250 | 210 | 270 | 330 | 390 | 450 |
| 500 | 360 | 420 | 480 | 540 | 600 |

determined using the estimating techniques described in Chapter II. The worth of this technique may be assessed by examining the achieved utilizations reported in Table II.

TABLE II

| Mean <br> Setup <br> Time | SYSTEM UTILIZATION IN PERCENT <br> Mean Job Size |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 91.7 | 90.8 | 86.1 | 88.0 | 93.0 |
| 50 | 92.1 | 91.7 | 92.3 | 90.6 | 92.5 |
| 100 | 90.2 | 90.4 | 93.7 | 88.9 | 92.4 |
| 250 | 85.2 | 92.4 | 92.4 | 90.6 | 94.4 |
| 500 | 88.3 | 89.1 | 92.4 | 93.9 | 94.4 |

The mean of the entries in this table is 91.1 percent. The extremes are 85.2 and 94.4 resulting in a range of 9.2 percent. Another measure of the worth of the estimating
procedure is displayed in Table III. The information in this table is the expected and achieved production in orders per day.

TABLE III
PRODUCTION IN ORDERS, PER DAY EXPECTED AND (ACTUAL)

| Mean Setup Time | Job Size Mix |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | (1) 100 | (.75)100 | (.5)100 | (.25)100 | (0) 100 |
|  | (0) 500 | (.25)500 | (.5)500 | (.75)500 | (1)500 |
| 0 | $\begin{gathered} 16.66 \\ (16.65) \end{gathered}$ | $\begin{gathered} 8.33 \\ (8.33) \end{gathered}$ | $\begin{gathered} 5.55 \\ (5.42) \end{gathered}$ | $\begin{gathered} 4.17 \\ (3.98) \end{gathered}$ | $\begin{gathered} 3.33 \\ (3.40) \end{gathered}$ |
| 50 | $\begin{gathered} 11.11 \\ (11.09) \end{gathered}$ | $\begin{gathered} 6.67 \\ (6.53) \end{gathered}$ | $\begin{gathered} 4.76 \\ (4.82) \end{gathered}$ | $\begin{gathered} 3.70 \\ (3.62) \end{gathered}$ | $\begin{gathered} 3.03 \\ (3.01) \end{gathered}$ |
| 100 | $\begin{gathered} 8.33 \\ (8.23) \end{gathered}$ | $\begin{gathered} 5.55 \\ (5.60) \end{gathered}$ | $\begin{gathered} 4.17 \\ (4.22) \end{gathered}$ | $\begin{gathered} 3.33 \\ (3.22) \end{gathered}$ | $\begin{gathered} 2.77 \\ (2.82) \end{gathered}$ |
| 250 | $\begin{gathered} 4.76 \\ (4.38) \end{gathered}$ | $\begin{gathered} 3.70 \\ (3.72) \end{gathered}$ | $\begin{gathered} 3.03 \\ (3.37) \end{gathered}$ | $\begin{gathered} 2.56 \\ (2.58) \end{gathered}$ | $\begin{gathered} 2.22 \\ (2.33) \end{gathered}$ |
| 500 | $\begin{gathered} 2.77 \\ (2.68) \end{gathered}$ | $\begin{gathered} 2.38 \\ (2.38) \end{gathered}$ | $\begin{gathered} 2.08 \\ (2.05) \end{gathered}$ | $\begin{gathered} 1.85 \\ (1.97) \end{gathered}$ | $\begin{gathered} 1.67 \\ (1.75) \end{gathered}$ |

To bring the system from idle to full operation as quickly as possible, 50 jobs are generated so as to enter the system simultaneously at the beginning of each run. Each run is permitted to continue 20 days before the process of collecting statistics begins.

Establishment of Due Dates

Since one measure of system performance, $E=D-A$, depends upon the due date established for each order processed through the system, it is important that due dates be set bias free with respect to order size. This is accom-
plished using the procedure described in Chapter II employing split statistics. It was decided to set the initial z at zero to improve visibility of results. No generality is lost since the effect of choosing $z$ is to alter the probability that a job will be early or late. Concern with this aspect of the measure seems appropriate only when there are costs associated with deviations from on time deliveries. Since this research is a study of the physical job shop system, rather than the economics of the system, taking $z=0$, which is equivalent to stating that the probability is 0.5 that a job will be completed on time, seems as good as any other choice.

The Experimental Runs

Each of the 25 primary experimental runs consisted of operating the system for 145 days. As previously noted, the first 20 days are employed to approach equilibrium performance. In addition to these primary runs, seven others were made, five for 520 days each, one for 900 days., and one for 1,800 days. The reasons for these seven runs will be discussed in Chapter IV.

The objective in every run is to measure or estimate equilibrium performance to increase comparability among runs. Most of the discussion so far in this chapter on the preliminaries of the design of the experiment has been to describe the choices made to achleve both visibility and comparability of results. The same conditions and
procedures are used for every run. The demand generator employs the same seed thus providing the same sequence of random numbers to control the arrival time and size of jobs entering the system. The remainder of this chapter is devoted to describing and explaining the conditions and intentions of the experiment.
Job Size Selection

The experiment consists of testing five mixes of two job sizes under five different conditions of setup time. As indicated, this produces 25 separate observations on each of the statistics of interest. Job sizes of 100 units and 500 units of homogeneous product are employed. They are combined in the following ways:

Mix
$1(1) 100+(0) 500=100$
$2(.75) 100+(.25) 500=200$
$3(.50) 100+(.50) 500=300$
$4(.25) 100+(.75) 500=400$
$5(0) 100+(1) 500=500$
The sum of these products are interpreted as follows: Mix 1 consists only of jobs of size 100; Mix 2 is 75 percent jobs of size 100 and 25 percent jobs of size 500 , etc. And, of course, the right hand side of the array contains the expected job size per mix.

The motivations for choosing jobs of sizes 100 and 500 are two in number. As is discussed in Chapter $V$, the
computer-program combination is very costly in scarce computer time. Diagnostic runs were initially accomplished with jobs of size 10 and size 100. The jobs of size 10 produced so many transactions as to make computer time requirements beyond that likely to be available. .... More important, jobs of size 10 produced such neat, "textbook" distributions of center flow time, total flow time, etc., as to be suspect. Further trials indicated the jobs of size 100 and jobs of another substantially larger size, 500, would minimize both objections. Finally, two sizes, rather than the 3,4 , or 5 of which the model is capable, were selected for the sake of simplicity. Of course, at least two sizes are required to produce mixes of sizes.

Completing the Design

It is possibie to design this experiment in a variety of ways. If one starts with the five job mixes just described, several options are possible. It seems appropriate to discuss some of these along with the design chosen to complete the job discussion of the conditions of the experiment and to introduce the discussion of the intentions.

One obvious and simple way to complete the design is to choose one common value of mean setup time at each center. This produces a results vector of five elements. A natural extension is to replicate each run several times with difo ferent sequences of demand caused by changing the random
number generator seed. Two or more levels of system utilization imposed along with these other conditions would appear to offer substantially correct design amenable to the statistical analysis of one system. However, this strategy and others similar to it are rejected in favor of one which provides the opportunity to acquire information about system reaction to changes in an important characteristic of job shops, setup time.

It was dechded to test each job mix under each of five different but common mean setup times at each of the ten centers in the system. For example, jobs of size 100 are tested under mean setup times of $0,50,100,250$, and 500 . In run 1, say, all center setup times are set to zero and all jobs are of size 100 in run 2, all setup times are 50 , and all jobs are size 100 , etc. The end results are arrays with 25 entries, 5 mixes by 5 setup times.

The advantages of this design are several. Even though this is a study of the physical aspects of the problem of job size, the ultimate interest will be in the economics associated with the results. Whether traditional inventory models apply to this work is of no special interest, but it is to be expected that the costs associated with inventory (in process) and setup time still will be appropriate. It has been shown by Little (26) that there are basically four measures of performance in the job shop; in-process inventory, utilization of centers, total flow time, and lateness. All of these are interrelated and associated with the cost
of operating the system. However, they provide only incomplete information when setup time and its associated costs are not included.

Does altering setup time in the selected fashion result in experimenting with one system or several? The question probably can be argued convincingly both ways. Earlier in this chapter, it is observed that several choices are required to delimit the experimental system. It is also noted in Chapter II that setup time is considered a function of the machine center and not the job or the sequence of jobs passing through it. Consequently, it is concluded that the experiment involves several systems, five to be exact, identical except with respect to setup time. This is taken to mean that there is no available rationale to permit statistical analysis of the joint results of one system. It means also that the design is in essence artificial (not possible with one real system) and is chosen only because of the overriding interest in seejng the results of operation under the different conditions of setup time Finally, on the matter of setup time, the magnitudes chosen correspond to multiples of processing time. It is of interest to note results when setup time is less than, approximately equal to, and greater than prom cessing time per order.

Hypotheses

There are generally two kinds of results to be expected
from experimentation of the kind being described; formulation of hypotheses and tests of hypotheses. Each of these is examined in turn.

Probably the most beneficial use of the model of the hypothetical system is in the formulation or discovery of apparently relevant questions during the course of the experimentation. Of course, some propositions occur to the researcher during the preliminary, problem definition phase of the research. Certainly this is true of the general question prompting the effort. Others arise during the diagnostic work with the model. More appear upon examination of the results of the experiment. It is clearly appropriate to test and to draw conclusions about those propositions arising in the problem definition phase. Here the propositions are stated in the absence of recognized order among the facts which may be at hand or the applicability of related theory with which the researcher may be familiar. In other words, questions translated into testable propositions at this point serve to direct the search for answers. It is considered important then, to set down propositions before the acts of testing or verification. Alternatively, it would be possible to "take credit" for propositions uncovered during the diagnostic and experimentation phases of the research; to accept as verified those relationships revealed in the course of the experimental runs. This approach is rejected as improper since further experimentation should be conducted with these
"revelations": carefully restated as testable propositions. All finite research efforts must terminate somewhere. Since this effort is not an exception, apparently relevant questions unearthed during the experiment will be discussed, restated as working hypotheses, some perhaps with tentative implications, and offered as propositions of possible worth for further research.

Propositions about the behavior of the job shop manufacturing system under the conditions specified for the experiment may be gleaned from the prior knowledge of the objects, attributes and relationships in the system established during the problem definition phase. Other sources of propositions are the disciplines and activities of industrial engineering, operations research, systems analysis, etc.; the prior work with job shop systems. Other plausible propositions have roots in recognizable bodies of theory such as queueing theory, network analysis, and inventory theory. It is not possible within the scope of the current effort to analyze the reactions of the system to all changes and reasonable propositions it is possible to contrive. It is necessary to be selective in what is chosen for study. As a consequence of this view, the discourse in Chapters IV and $V$ pertaining to the analysis of the results and the conclusions to be drawn will be restricted to the following questions:

1. What is the effect or order size on the idle time in the system?
2. What is the effect of order size on the total flow time of orders through the system?
3. What is the effect of order size on the measure, $E=D-A$ ?
4. What is the effect of order size on the various queues in the system?

It remains to stipulate that the conventional null hypothesis is taken for each of these questions. Where appropriate, statistical hypotheses are stated and tested.

In this experiment, the null hypotheses are of the form:

$$
\begin{array}{ll}
H_{0}: & \Sigma \phi_{j}=0 \\
H_{1}: & \Sigma \phi_{j} \neq 0 .
\end{array}
$$

The interpretation is as follows: The null hypotheses, $H_{0}$, imply that there are no differences in the measured attributes of idleness, time in the system, lateness, and waiting lines, caused by job size. The alternate hypotheses, $H_{1}$, imply that job size does indeed cause some significant differences. It is hoped, of course, that some null hypotheses will be accepted and some rejected.

```
Statistical Models
```

Some of the experimental data are investigated by employing a fixed effects analysis of variance model, ANOVA. In the fixed model a difference in mean response at a certain level of significance is detected by an F ratio of
the mean square of the columns (in this study) to the residual mean square. Note that all ANOVA are fixedeffects, 2-way, one observation per cell.

The model for this situation is a statement of linear treatment effects as follows:

$$
x_{i j}=\mu+\gamma_{i}+\phi_{j}+\varepsilon_{i j}: \begin{aligned}
& i=1,2,-\cdots, r \\
& j=1,2,-\cdots, r
\end{aligned}
$$

where $\mu$ is the general mean, and $\varepsilon_{i j}$ are the experimental errors which are assumed to be normally distributed, each with mean zero and variance $\sigma^{2}$.

In the fixed effects model with one observation per cell:
$\gamma_{i}$ is the effect of adding the th row treatment

$$
\sum_{i=1}^{r} \gamma_{i}=0
$$

$\phi_{j}$ is the effect of adding the $j$ th fixed column treatment

$$
\sum_{j=1}^{c} \phi_{j}=0
$$

A second model is employed to generate the coefficient of correlation $r$ where

$$
r= \pm \sqrt{1-\frac{\Sigma\left(y-y^{\prime}\right)^{2}}{\Sigma(y-y)^{2}}}
$$

In words, we compare the sum of the squares of the vertical deviations from the least-squares line with the sum of the squares of the deviations of the $y^{\prime}$ 's from their mean. The proper hypothesis in this situation is

$$
\begin{aligned}
& H_{0}: \rho=0 \\
& H_{1}: \rho \neq 0 .
\end{aligned}
$$

The test for significance may be summarized as follows:

$$
\text { if the }|r|>\left|r_{\alpha / 2}\right| \text {, reject } H_{0} \text {. }
$$

Choice of Significance Level

The five percent level of significance is chosen for the statistical analyses because it is commonly used and extensively tabulated for Snedecor's F.

## CHAPTER IV

## ANALYSIS

The goal of this chapter is a lucid and detailed description and analysis of the results of the experiment. The questions posed for this investigation will be treated in the order listed in Chapter III, namely: idleness, flow time, lateness, and waiting time。 Certain sections are devoted to observations not properly a part of the analysis of the four primary questions.

## Idleness in the System

If any center in the system is not engaged in physically altering a unit of product, it is said to be idle。 Idleness, then, is the condition of not doing work. The attributes of idleness chosen for examination are the parameters and shapes of the distributions of idle time occurring at each center in the system.

In the language of Chapter II, and referring to Figure 1, page 14, if block $P$ is not occupied, the center, T, is idle. If block $S$ is occupied, $P^{\prime}$ s idleness is caused by the occasion of setup. It makes no difference if block $Q$ is empty or full. If $Q, S$, and $P$ are all empty, however, idleness is not caused by setup but by absence of work at
the center.
It will be recalled from Chapter III that the arrival rate of jobs was determined to achieve a 90 percent utilization of each center in the job shop system. This is equivalent to stating that idleness caused by the absence of work at any center is 10 percent. Of course, 90 percent utilization was not achieved in every case, hence, neither was 10 percent $i d j e n e s s, ~ b e c a u s e ~ o f ~ t h e ~ a b s e n c e ~ o f ~ w o r k . ~$ What was achieved is displayed in Table IV.

TABLE IV.
MEAN IDLENESS IN PERCENT CAUSED bY ABSENCE OF WORK

| Mean <br> Setup | Mean Job Size |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Time | 100 | 200 | 300 | 400 | 500 |
| 0 | 8.3 | 9.2 | 13.9 | 12.0 | 7.0 |
| 50 | 7.9 | 8.3 | 7.7 | 9.4 | 7.5 |
| 100 | 9.8 | 9.6 | 6.3 | 11.1 | 7.6 |
| 250 | 14.8 | 7.6 | 7.6 | 9.4 | 5.6 |
| 500 | 11.7 | 10.9 | 7.6 | 6.1 | 5.6 |

Since the mean idleness caused by the absence of work is fixed by the choice of the utilization rate, it is not of special interest in this study.

Idleness caused by setup at a machine center is exactly equal to the time required for each setup multiplied by the number of setups. In symbols

$$
I_{S}(j)=n E\left[S_{j}\right]
$$

In different words, the total idle time, $I_{s}$, because of setup at center $j$ is equal to the product of the number of times setup occurred, $n$, and the expected value of setup at center $j$. Table $V$ contains the mean idle time caused. by setup in each test. Easily seen is the well understood fact that setup time causes loss of production in direct proportion to the product of its occurrence and magnitude.

TABLE V
MEAN IDLENESS IN PERCENT CAUSED BY SETUP TIME

| Mean <br> Setup | Mean |  |  |  |  |  | Size |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Time | 100 | 200 | 300 | 400 | 500 |  |  |  |
| 0 | 91.7 | 90.8 | 86.1 | 88.0 | 93.0 |  |  |  |
| 50 | 30.7 | 18.2 | 13.4 | 10.0 | 8.4 |  |  |  |
| 100 | 45.2 | 49.5 | 23.6 | 22.8 | 15.5 |  |  |  |
| 250 | 60.8 | 51.3 | 41.5 | 35.3 | 31.5 |  |  |  |
| 500 | 73.6 | 65.4 | 57.6 | 52.6 | 47.2 |  |  |  |

Additionally, setup time defines a lower bound on idle time such that no incident of idle time can be less than the smallest possible setup time. For example: It will be recalled that $S$ is drawn from the uniform distribution with range $E[S] \pm(0.1) S$. Suppose $E[S]=100$. The minimum value S can assume is 90 , and 90 , then, is also, the minimum possible magnitude of idle time. This effect of setup time on idle time is an unsought consequence of the research and does not appear to bear directly upon the questions addressed.

As implied earlier in this section, for a given common setup time and fixed system utilization rate, there is no effect of order size on the mean idle time caused by the absence of work at centers in the system. On the other hand; mean idle time resulting from setup decreases as order size increases. Since this is true under the condition of common setup time, it must follow that the true effect of increasing the mean order size is to reduce the number of setups. This is not an unexpected result. Table VI compares five ratios of the mean number of setups at each center to the units of product processed by the system.

TABLE VI
RATIO OF MEAN NUMBER OF SETUPS PER CENTER
TO SYSTEM PRODUCTION IN JOBS

| Mean Job Size |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Setup | 100 | 200 | 300 | 400 | 500 |
| 500 | .0047 | .0024 | .0015 | .0011 | .0009 |

That these ratios decrease as order size increases substantiates the previous conclusion. Ratios, rather than absolute values, were employed because of unequal'production.

Order size has an effect on the dispersion in the idle time distribution and the larger portion of this effect is on idle time generated because of the absence of work. This must be the case since idle time caused by setup also has an upper bound. If $E[S]=100$, then the maximum idle time caused by setup is 110 units for each job passing through a
center. Hence, it follows that $90<I<110$ from this and the previous example on the lower bound of $I$. Continuing with the case where $E[S]=100$, the mean range of idle time tends to increase with the increase in job size. The standard deviation increases also, but the range appears to be more descriptive of the nature of the dispersion as will be discussed. Below are the ranges of idle times for each mean job size when $E[S]=100$.

Job Size
100 200

300
400
500

Range (days)
1.21
2.41
1.91

The distributions of idle times at all centers in all runs with $S$ > 0 perhaps are described best by taking advantage of the way in which the statistics are recorded. An example of this is contained in the 20 series table in Appendix C. Statistics are recorded in increments of 100 clock units (tenths of days). The number of times idleness occurs such that its magnitude lies between, say, 101 and 200, is recorded in class interval 200. The result is a histogram, Figure 3. The magnitude of idle time by class is recorded on the abscissa and frequency of the magnitude on the ordinate. Viewed in this artificial way, the distributions of idle time are essentially 2-valued. Figure 3 displays the distribution of idle time for center 3 , with


Figure 3. Distribution of Observed Idle. Time, Center 3, Run 100-100
$E[S]=100, E[P]=100$. Actually, in this case, 95.37 percent of idle time 1 ies between 90 and 200 clock units (between. 09 and .2 days). As job size increases, this percentage increases until for jobs of size 500 it is 97.88 . The net effect of increasing the mean job size passing through the system seems to be to increase the concentration of idle time near the mean of the population and, at the same time, to create small numbers of increasingly longer periods of idleness. This explains the preference for the range as a measure of dispersion.

It is concluded that job size:

1. does not affect the mean idle time in the system caused by absence of work at centers in the system,
2. does affect mean idle time caused by setup requirements at the centers in the systems by altering the number of setups required,
3. does affect the dispersion of idle time distributions at centers in the system by creating a small number of increasingly long periods of idleness as job size increases.

Consequently, the null hypothesis, that job size does not affect idle time, is not accepted.

Speculation on Idle Time

The examination of the effect of order size on the idle time in a job shop manufacturing system prompts some observations about idle time not appropriately a part of the previous discussion. If it is assumed that reducing idle time is a preferred course of action, then it is important to suggest ways in which this might be accomplished. The point of departure for the discussion of one possible way is the system employed in this study.

A moments reflection will substantiate that, in a system with $S$, 0 , the frequency of occurrence of idle time at each center is equal to the number of orders passing through a center. Further, idle time caused by absence of work invariably precedes idleness caused by setup. The implications
of these conditions are fairly obvious. First, there appears to be an opportunity to reduce idle time caused by setup. Second, the way to accomplish this is to arrange for setup to occur concurrently with the absence of work. For ease of discussion, idle time per job, I, is the sum of idle time caused by absence of work, $A$, and idle time caused by setup, $S$. Then $I=A+S$ not only describes the amount of idle time associated with a given order but also the proper chronological order of $A$ and $S$.

As previously discussed $S>0$. However, $A \geq 0$. As a matter of fact, $A=0$ is the rule rather than the exception. Of course, $A=0$ is equivalent to stating that the order about to be setup is already in the queue where the work is to be performed. In this study, $A>0$ occurred about $15 \%$ of the time. Further, $A$ \& $S$ occurred more frequently by far than $A \geq S$. Hence, the concurrence of $A$ and $S$ is not expected to be complete and the reduction in $I$ is expected to be small, especially in systems with high utilization。

The development of this proposition would be incomplete without offering some ideas about the kinds of control information required to achieve partial concurrence of $A$ and $S$ in the job shop manufacturing system。 If it is assumed that the empty center will begin work (setup) on the first job to arrive, then it remains only to determine which job among the other centers in the system (or dispatching) will arrive next and the specifications of the required operation.

For clarification, consider a job shop system of three
centers. Center 3 in this system is idle; Centers 1 and 2 are occupied. If neither job in Center 1 or 2 is scheduled for 3 , there is no action to be taken in Center 3 ... If one of the two jobs is scheduled for Center 3, then it is an easy matter to begin to prepare Center 3 for that job. If both jobs at Centers 1 and 2 are scheduled for Center 3 it is necessary to determine which will be completed first. When this is determined, setup at Center 3 may begin. If it begins before the job is through at the preceding center, part or all of the idleness because of setup time may be saved. Suppose it is determined that the job at Center 2 will be completed before the job at Center 1。 If Center 3 is setup and ready to begin processing the job from Center 2 before it leaves Center 2, then all of the idleness due to setup is saved. If the setup is half complete before the job from Center 2 arrives, then half of the idleness due to setup is saved. Permitting a downstream center to prepare for jobs that have yet to arrive reduces total idleness at the center by the amount of setup that can be completed before the arrival of the jobs. It is this idea, then, which has been labeled "the partial concurrence of $A$ and $S "$.

The concept of partial concurrence of $A$ and $S$ is not new. The advantages of parallel, simultaneous, or overlapping operations seem to be well understood in other forms of activity. An unlikely analogy comes from the game of contract bridge where the declarer often has to contrive a way to combine two losing tricks into one by playing the
losing cards on the same trick.
The literature search preceding this dissertation did not reveal any treatment of the proposition of reducing idle time in the job shop manufacturing system by the means of partial concurrence. It appears, therefore, to be a worthwhile subject for further investigation.

## Order Flow Time

The next proposition to be considered is that job size has no effect upon the time it takes orders to traverse the system. As with idle time the attributes of flow time are the parameters and shapes of the distributions of flow times generated at the centers in the system and the distribution of flow time through the entire system, or total flow time.

The subject of total flow time is considered first.
Here there are three propositions about expected flow time:

1. mean job size does not affect expected total flow time,
2. mean job size does not affect the expected total flow time of jobs of size 100 ,
3. mean job size does not affect the expected total flow time of jobs of size 500. Each of these propositions is tested by ANOVA as described in Chapter III.

As may be seen from examining Table VII, the test of: the first proposition seems almost trivial. still it is of some interest to see a statistical conformation of the

TABLE VII
AVERAGE TOTAL FLOW TIME IN DAYS/ORDER

| Mean <br> Setup <br> Time | 100 | Mean Job Size |  |  |  |  |  |
| ---: | ---: | ---: | ---: | ---: | ---: | :---: | :---: |
| 0 | 3.150 | 10.221 | 7.630 | 10.239 | 18.209 |  |  |
| 50 | 4.763 | 10.280 | 15.124 | 15.407 | 16.870 |  |  |
| 100 | 5.468 | 12.853 | 17.613 | 12.202 | 18.301 |  |  |
| 250 | 7.047 | 15.917 | 17.854 | 17.359 | 23.123 |  |  |
| 500 | 14.749 | 17.710 | 22.020 | 26.524 | 27.248 |  |  |

anticipated outcomes and to compare the effects of job size and setup time on total flow time. Table VIII contains the results of the ANOVA calculations to test the proposition that mean job size does not affect expected total flow time.

TABLE VIII
ANOVA, TOTAL FLOW TIME

|  | Degrees of <br> Freedom | Sum of <br> Squares | Mean <br> Square | Test |
| :--- | :---: | :---: | :---: | :---: |
| Tource | 24 | 983.442 |  |  |
| Job | 4 | 403.817 | 100.953 | $22.489 \%$ |
| Setup | 4 | 507.798 | 126.949 | $28.280 \%$ |
| Residual | 16 | 71.827 | 4.489 |  |
| Fo.05, 4, 16 | $=3.01$ | *Reject $H_{0}$ |  |  |

The raw data for these calculations is taken from Table VII above. Of course, the null hypothesis is rejected and it is concluded that job size does alter expected total flow time. This is considered a natural result. As jobs
increase in size, they require more work, hence, more time at each center and in the system.

The next two propositions require the use of responses in Table IX. This table shows total flow time by job size rather than by expected job size. Additionally, all responses are not employed in the analyses. The first row is deleted. Its purpose, that of serving as a means of observing certain aspects of model performance, is fulfilled.
table IX
AVERAGE TOTAL FLOW TIME PER ORDER by SIZE

| Mean Setup | Mean Job Size |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Iime | 100 |  | 200 |  | 300 |  | 400 |  | 500 |  |
|  | 100 | 500 | 100 | 500 | 100 | 500 | 100 | 500 | 100 | 500 |
| 0 | 3.2 | 0 | 9.3 | 12.9 | 6.3 | 9.0 | 7.9 | 11.0 | 0 | 18.2 |
| 50 | 4.8 | 0 | 9.4 | 12.7 | 13.8 | 16.5 | 13.4 | 16.0 | 0 | 16.9 |
| 100 | 5.5 | 0 | 12.0 | 15.9 | 15.5 | 19.6 | 9.5 | 13.2 | 0 | 18.3 |
| 250 | 7.0 | 0 | 14.9 | 19.0 | 15.6 | 20.1 | 19.0 | 16.9 | 0 | 23.1 |
| 500 | 14.7 | 0 | 17.8 | 17.4 | 21.9 | 22.1 | 27.2 | 26.3 | 0 | 27.2 |

The responses under the conditions of zero setup time are not considered comparable to the other row responses.

Additionally, column 5 is deleted for the test of the second proposition and column 1 is deleted from the test of the third proposition. Tables $X$ and $X I$ display the results of the ANOVA calculations for both of these tests. The null hypothesis is rejected in both cases and it is concluded that expected job size does alter the total flow time of the two individual job sizes.

TABLE X
ANOVA, TOTAL FLOW TIME, SIZE 100

|  | Degrees of <br> Freedom | Sum of <br> Squares | Mean <br> Square | Test |
| :--- | :---: | :---: | :---: | :---: |
| Source | 15 | 533.0 |  |  |
| Total | 3 | 258.7 | 86.2 | $10.3 \%$ |
| Job | 3 | 199.0 | 66.3 | 7.9 |
| Setup | 9 | 75.3 | 8.4 |  |

$$
F_{0.05}, 3,9=3.86 \quad * \text { Reject } H_{0}
$$

It is interesting to note from Table XI that setup does not significantly alter flow time for jobs of size 500. The explanation is believed to lie in the fact that for the larger job, setup time is relatively smaller; e.g., there is no case tested where job size is smaller than setup.

TABLE XI
ANOVA, TOTAL FLOW TIME, SIZE 500

|  | Degrees of <br> Freedom | Sum of <br> Squares | Mean <br> Square | Test |
| :--- | :---: | :---: | :---: | :---: |
| Source | 15 | 257 |  |  |
| Total | 3 | 142.5 | 47.5 | $7.3 *$ |
| Job | 3 | 55.6 | 18.5 | 2.8 |
| Setup | 9 | 58.9 | 6.5 |  |

$F_{0.05,3,9}=3.86$
*Reject $\mathrm{H}_{0}$

All three statistical analyses of mean total flow time produce a small residual mean square. Residual mean square is composed of interaction between setup and job size as well as error mean square. Consequently, the small magnitudes suggest the absence of interaction. Conformation of this would require replication of the experiments and use of an expanded linear model to identify mean response because of interaction.

The dispersion in the total flow distributions increases with both job size and setup time as may be seen in Table XII. Here are recorded the magnitudes of one standard deviation in days from each of the 25 total flow time distributions. Another approach to examining the dispersion

TABLE XII
MAGNITUDE IN DAYS OF ONE STANDARD DEVIATION IN TOTAL FLOW TIME

| Mean <br> Setup <br> Time | Mean Job Size |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 0 | 100 | 200 | 300 | 400 | 500 |  |
| 50 | 2.8 | 6.0 | 5.2 | 3.0 | 9.1 |  |
| 100 | 3.0 | 7.1 | 9.7 | 7.5 | 9.6 |  |
| 250 | 4.3 | 8.7 | 9.5 | 9.4 | 11.4 |  |
| 500 | 8.7 | 9.9 | 11.8 | 14.1 | 13.7 |  |

under the various experimental conditions is to compare the variance or standard deviation along one common path through the system. This is done in Table XIII by computing the standard deviation along the path through the system which
contains each center only once. This path can occur in 10! or $3,628,800$ ways, roughly 0.3 percent of all possible paths.

TABLE XIII
STANDARD DEVIATION IN DAYS OF A TEN CENTER PATH

| Mean <br> Setup | Mean Job Size |  |  |  |  |  |
| ---: | ---: | ---: | ---: | ---: | ---: | :---: |
| Time | 100 | 200 | 300 | 400 | 500 |  |
| 0 | 2.238 | 8.083 | 5.233 | 4.460 | 10.817 |  |
| 50 | 3.187 | 6.985 | 9.412 | 9.059 | 9.591 |  |
| 100 | 3.520 | 8.007 | 10.271 | 7.434 | 10.906 |  |
| 250 | 4.290 | 8.590 | 10.706 | 9.706 | 11.200 |  |
| 500 | 8.748 | 9.372 | 10.530 | 11.132 | 11.684 |  |

Finally, the shape of both the center flow time and total flow time distributions appear only slightly changed by changing job size. As observed in Chapter III, the center flow time distributions are Poisson-like. This characteristic remained essentially unchanged during all runs. The total flow time distributions are best described by the uniform distribution, although there is a slight tailing-off at the upper magnitudes of flow time。 The same uniform character holds for the total flow time distributions recorded for individual job sizes. These statistics were maintained when mixes of jobs were fed through the system. (Mixes are columns 200, 300, and 400 in all tables using this type identification.)

The analysis of flow time leads to the conclusion that job size does alter the means and standard deviations of the flow time distributions, but leaves the shape of the distributions essentially undistrubred.

## Late Delivery of Orders

The third proposition to test is that job size does not affect the value of the measure, $E=D-A$, where $D$ is the due date of the order and $A$ is the completion date. Positive values of $E$ indicate that an order is early, negative values of $E$ indicate that it is late. It will be recalled that initial $z=0$ in this experiment so that the E[E] = 0 . This means, of course, the responses recorded in Tables XIV, XV, and XVI should be near zero or of like magnitudes among column entries to conclude that job size does not alter E.

TABLE XIV
D-A IN DAYS, POSITIVE ENTRIES ARE DAYS EARLY

| Mean <br> Setup <br> Time | Mean Job Size |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | 100 | 200 | 300 | 400 | 500 |
| 0 | .176 | -3.203 | . 736 | . 251 | -4.309 |
| 50 | .171 | -1.985 | -2.977 | -2.954 | -2.802 |
| 100 | -0.152 | -1.541 | -5.361 | - . 189 | -3.976 |
| 250 | . 226 | -2.967 | -4.056 | -3.106 | -6.581 |
| 500 | $-1.169$ | $-2.656$ | $-6.156$ | -8.892 | -9.464 |

The means, standard deviations, and shapes of the distributions of delivery times are the attributes of interest in this analysis.

Table XIV is the array of mean $E$ recorded at the end (145 days) of the experimental runs. That both increases in job size and setup alter E, tend to increase late deliveries, is fairly obvious.

Table XV, shows mean $E$ at a point of equal production in all cases (approximately 40,000 units of product). Again the difference in the mean response of $E$ is pronounced.

TABLE XV

> D-A IN DAYS, EQUAL PRODUCTION IN UNITS, POSITIVE ENTRIES ARE DAYS EARLY

| Mean <br> Setup | Mean Job Size |  |  |  |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | ---: |
| Time | 100 | 200 | 300 |  | 400 | 500 |
| 0 | -.378 | -1.031 | -.2 .385 | -3.368 | -10.973 |  |
| 50 | -.112 | -1.949 | -4.796 | -5.543 | -11.452 |  |
| 100 | -.057 | -3.598 | -5.128 | -6.615 | -12.014 |  |
| 250 | .438 | -4.801 | -7.785 | -10.255 | -14.109 |  |
| 500 | .321 | -4.619 | -11.318 | -17.503 | -19.937 |  |

Table XVI, displays the same information as the previous table except at a point of equal production in jobs completed (approximately 220). There is again no change in the marked affect of job size on late deliveries. Note the value of the periodic status report as an analytical tool. Without it there would have been no way to compare the
values of E except with terminal statistics; i.e., at the end of 145 days.

TABLE XVI
D-A IN DAYS, EQUAL ORDER PRODUCTION, POSITIVE
ENTRIES ARE DAYS EARLY

| Mean <br> Setup <br> Time | Mean Job Size |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | 100 | 200 | 300 | 400 | 500 |
| 0 | - . 224 | $-1.031$ | - . 446 | - . 991 | -5.744 |
| 50 | - . 332 | -1.819 | -3.281 | -4.460 | -4.922 |
| . 100 | - . 664 | -3.598 | $-4.356$ | -2.499 | -5.396 |
| 250 | . 408 | -4.145 | -5.456 | -4.885 | -7.960 |
| 500 | -2.668 | -4.408 | $-6.723$ | -9.355 | -9.464 |

A typical distribution of $E$ is shown in Figure 4. Increase in job size does not alter the shape of this distribution except to increase the length of the tails in both directions but mainly in the direction of late deliveries. In Chapter III interest was expressed in achieving comparable results to those reported by Fabrycky and Shamblin (20) in their test of the probability based sequencing algorithm。 Their work shows the distribution of E skewed towards early delivery. The results of this experiment show the distribution E skewed toward late delivery. No other differences are apparent.

The results clearly indicate the rejection of $H_{0}$. Increase in job size contributes significantly to lateness. However, this conclusion is offered with considerable


Figure 4. Typical Distribution of $E=D-A$
reservation as will be discussed below.

Reservations on Lateness

E is a relative measure. Its magnitude depends upon the predetermined due date, D. D is a changing standard against which to measure since its value is partly a function of the mean performance of the system. In other words, a feedback loop is employed to adjust the computation of $D$ to correspond to the current state of the job shop manufacturing system. This is accomplished, of course, taking into account the differences in expected setup and processing times for the two different job sizes, thus
removing bias because of job size and setup.
One of the necessary conditions for comparability of results in situations like this is the condition of equilibrium.

So far as the systems and jobs flowing through them are concerned, the general state of the process, there is every reason to believe that equilibrium conditions exist. The conditions observed are best described to be like statistical control, a stable mean with random fluctuations about the mean as in Figure 5, However, the same cannot be said, in all cases, about the generation of $E$.


Time
Figure 5. Statistical Control

Seven additional simulations were performed repeating the runs with $E[S]=500$. Mean job sizes through 400 were run once at 520 days each. Job size 500 was run three times at 500,900 , and 1,800 days. None of these runs produced
any results indicating that $E[E]$ had stabilized. Consequently, the results of the previous section are, at a minimum, suspect.

Even partial failure is not without its reward, however. It turns out that E[E] did stabilize in at least one run of smaller job size and less setup time. ...Inis particular run produced 2000 completed jobs. The final run of 1,800 days ( 7.2 years) produced 3000 completed jobs and had yet to achieve stability. This comparison suggests that increasing job size or setup time or both have a marked impact on the rate of stabilization of E[E]. It might prove valuable to test the proposition that as the ratio of job size to system capacity increases, due date based sequencing algorithms tend to lose their efficiency. From the practical point of view, it is difficult to visualize a job shop like system in operation for more than seven years without a significant change in some of its characteristics.

## Waiting Time

The final proposition is that job size does not alter the times a job waits in the various queues in the system. In addition to the parameters and shapes of the distribution of waiting time at the centers in the job shop manufacturing system, it is of interest to discuss the jobs which do not have to wait.

Table XVII shows the mean waiting time per job per center for each of the 25 tests. Statistical analysis of

TABLE XVII
mean waiting time per job per center IN DAYS

| Mean <br> Setup <br> Time | Mean Job Size |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | 100 | 200 | 300 | 400 | 500 |
| 0 | . 462 | 1.695 | . 959 | 1.451 | 2.726 |
| 50 | . 473 | 1.625 | 2.425 | 2.172 | 2.365 |
| 100 | . 790 | 1.925 | 2.487 | 1.680 | 2.629 |
| 250 | . 878 | 2.313 | 2.549 | 2.355 | 2.846 |
| 500 | 1.737 | 2.289 | 2.654 | 3.069 | 3.193 |

these responses, Table XVIII, requires that the null hypothesis be rejected. Variance in the waiting time distributions (not displayed) increased in the same manner as the mean waiting time. The effect of increase in job sizes

TABLE XVIII
ANOVA, MEAN WAITING TIME PER JOB PER CENTER**

|  | Degrees of <br> Freedom | Sum of <br> Squares | Mean <br> Square | Test |
| :--- | :---: | :---: | :---: | :---: |
| Total | 24 | 149,270 |  |  |
| Job | 4 | 36,533 | $9,133.25$ | $8.3 *$ |
| Setup | 4 | 95,158 | $23,789.50$ | 21.7 |
| Residua1 | 16 | 17,579 | $1,098.70$ |  |

$$
F_{0.05} ; 4,16=3.01
$$

on the shape of the distribution of waiting time is slight. The tail of the distribution is lengthened and rate of change of slope is slightly reduced.

Table XIX shows the percentage of jobs which, on the average, did not have to wait. The correlation between these figures and corresponding system utilization shown in Table II is obvious. As system utilization decreases, the

TABLE XIX
MEAN PERCENT OF JOBS PER CENTER WITH ZERO WAITING TIME

| Mean <br> Setup <br> Sime | Mean Job Size |  |  |  |  |  |
| :---: | ---: | ---: | ---: | ---: | ---: | :---: |
| 0 | 100 | 200 | 300 | 400 | 500 |  |
| 50 | 7.1 | 9.8 | 14.1 | 13.4 | 7.5 |  |
| 100 | 10.5 | 10.2 | 8.6 | 11.9 | 9.1 |  |
| 250 | 15.6 | 9.2 | 7.7 | 11.1 | 7.0 |  |
| 500 | 13.2 | 12.3 | 8.0 | 8.0 | 7.0 |  |

mean percentage of jobs receiving service without waiting increases. The computations are not shown, but the correlation between these two types of response is high, $r=-.84 . \quad$ This relationship leads to interest in another, namely the mean percentage of jobs not waiting and the corresponding production time. The coefficient of correlation of these data is calculated below.

$$
r=\frac{n(\Sigma x y)-(\Sigma x)(\Sigma y)}{\sqrt{n\left(\Sigma x^{2}\right)-(\Sigma x)^{2}} \sqrt{n\left(\Sigma y^{2}\right)-(\Sigma y)^{2}}}
$$

$$
\begin{aligned}
r & =\frac{198,620-208,190}{(396.2)(26.24)} \\
& =-0.52 .
\end{aligned}
$$

Employing the standard critical value of $r$, assuming the $x$ 's as constants and the y's as normally distributed with common variance $\sigma^{2}$, we may reject $H_{0}: \rho=0$ and accept $H_{1}: \rho \neq 0$ based upon $r=-0.52<r .025=-0.444$, for a sample size of 20 . Note that row 1 was deleted from this calculation since under perfect conditions, $r=-1.0$ for the responses under conditions of zero setup.

## CHAPTER V

## SUMMARY AND CONCLUSIONS

This chapter is composed of five sections. The first is a brief summary of the research effort. The second contains the conclusions reached. The third offers proposals for further study. The fourth acknowledges possible sources of errors and the fifth treats some practical considerations involving the computer and program employed in this study.

## Summary

This investigation treated the general question of the effect of order size on the operation of a job shop manufacturing system. Chapter II described the computerized model built for the research. Chapter III was the exercise of designing the experiment to produce reasonable, visible and comparable results. . The need to make careful choices was emphasized. Hopefully, any errors in this work are the result of making wrong choices rather than overlooking situations where choices should have been made. Chapter IV deals with the analysis of the four propositions under investigation.

## Conclusions

The relationships between job size and job shop operation as derived in this dissertation indicate that increases in job sizes:

1. increases production in the job shop by reducing the incident of setup,
2. increases the total flow time of orders through the system in proportion to the increase in job size,
3. enlarges the means and variances of all time related distributions in the system,
4. do not materially alter the shapes of the time related distributions.

The major worth of this research effort appears to be in four areas. First, the question addressed has not, in the knowledge of the researcher, been treated before. That it has now been asked and partially answered should be a step forward. Second, the work serves to confirm existing theory, not by repeating previous experiments, but by the employment of a refined technique of splitting center flow time into component parts of process, setup and queue time. This categorization and others are articulated in the literature but there is no evidence that they have been employed in models of systems. Third, the estimating technique developed to set mean arrival times, while simple enough, appears to be new and useful ableit limited in


#### Abstract

application. Finally, the concept of reducing system idle time by causing essentially two kinds of idle time to coalesce seems important.


## Future Research

Probably the most interesting proposition for future study is the possibility of devising a repetitive decision rule to reduce idle time through achieving partial concurrence of idle time caused by absence of work and idle time caused by setup time。

A second question to be addressed is the feasibility of due date based sequencing algorithms in low production situations as discussed in Chapter IV.

Due dates themselves deserve additional attention. There is little evidence in the literature to indicate research on this subject. There appears to be a need to objectively examine several alternative ways of assigning due dates to determine their relative merits. It is suggested that any examination of due dates should be accomplished by taking into account the economics associated with deviations from on time deliveries.

It would be worthwhile to reproduce this study with minor adjustment to explore more mixes of jobs and more basic job sizes. The purpose would be to discriminate more finely the differences which occur and to introduce replication to test for interaction.

## Possible Error Sources

As discussed in Chapter IV, the probability of an error in the results of the analysis of lateness remains because of the apparent inability to achieve steady state conditions for $E[E]$ in the low production situations.

Another possible source of error is the assumption of normality of the distribution of total flow time of jobs through the system. As noted in Chapter IV, total flow time distributions were more nearly uniform.

Statistics recorded by class interval tend to conceal the true shape of the distribution of variables. While care was exercised, it is possible that error is present.

## The Computer and The Program

This subject is saved until last because it bears more on the possible future work of others than on this research.

Considerable care was employed in constructing the computer program so that it might be used by others. Examination of Appendix $D$ will show that the program is carefully annotated as to the function of all routines.

It is now necessary to recommend that it not be used. The programming language, GPSS II with FORTRAN, when combined with the UNIVAC 1107, on which this work was accomplished, is painfully slow. The diagnostic and experimental runs for this study consumed more than 50 hours computer running time. Fortunately, GPSS II is now available with FORTRAN. There is also a routine to convert this study's
program to GPSS III. If converted and rerun on, say, the IBM 7090, the running time should be less than 10 hours - a substantial savings. There are also basic errors in the version of GPSS II employed. The version is called EXEC II. The two errors causing the most difficulty are the inconsistent use of the relative and absolute clocks and the failure to provide the means to produce both weighted and unweighted statistics. Both capabilities are described in the programming manual but are absent in the EXEC II version of GPSS II。

The problem of the relative and absolute clocks resulted in considerable difficulty in delaying the collection of statistics until equilibrium conditions were achieved. As a result, it was necessary to bypass this feature of GPSS II and develop a FORTRAN subroutine to recycle the summary statistics.

The second program error was never corrected. Histograms are either weighted or unweighted, but not both for any given variable. In this study, it would have been an advantage to be able to compare weighted to unweighted entries in class intervals because of the mixes of jobs. The existence of the mixes made it difficult to relate the numbers of jobs to the numbers of units of product within corresponding intervals.

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## APPENDIX A

## TRACE REPORT

The trace report is generated by introducing NTYPE(3) job into the system. The trace report starts when the NTYPE(3) job enters the system and is suppressed when it exits the system. The reports are printed with the same frequency set for the periodic status report (explained in Appendix $B)$, but include all relevant times in the interval between reports. Trace reports contain nine columns of differentiated information in plain language, depending upon the action being taken. No explanations of column entries are considered necessary.

| *CHK 2 Job | 42 | IN | QUE | 2 | CLOCK | 20169 | D-C-10168 | $z=-7.81$ | 2 IN CTR=-19.43 | NEXT Z $=-20.90$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| *CHK 2 Job | 4 | IN | QUE | 9 | CLOCK | 20169 | D-C-10168 | $z=-4.89$ | 2 IN CTR=150.00 | NEXT $\mathbf{z =}=-7.31$ |  |  |  |  |  |
| *SETUP JOB | 35 | IN | CTR | 9 | CLOCK | 20169 | D-C-10168 | $z=-7.31$ | 5 MEAN 500 | SPREAD 50 |  |  |  |  |  |
| *CHK Z JOB | 42 | IN | QUE | 2 | CLOCK | 20170 | 0-C-10169 | $z=-7.81$ | 2 IN CTR=-20.91 | NEXT Z $=-20.90$ |  |  |  |  |  |
| *SEND JOB | 46 | TO Qu | QUE | 2 | CLOCK | 20210 | D-C-10209 | $z=-7.29$ | T MEAN 1848.65 | ST.DEV 1023.05 | Q | MEAN | 567.00 | ST.DEV | . 00 |
| *CHK 2 JOB | 46 | IN | QUE | 2 | CLOCK | 20210 | D-C-10209 | $z=-7.29$ | 2 IN CTR=-20.91 | NEXT $Z=-20.86$ |  |  |  |  |  |
| *PROC. JOB | 35 | IN | CTR | 9 | CLOCK | 20715 | D-C-10714 | $z=-7.31$ | P MEAN 100 | SPREAD 10 |  |  |  |  |  |
| *EXIT JOB | 35 | FROM | M SHOP |  | CLOCK | 20816 | D-A-10815 | $z=-7.31$ | SIZE 100. DUE | 10001 |  |  |  |  |  |
| *CHK 2 JOB | 4 | IN Q | QUE | 9 | CLOCK | 20816 | D-C-10815 | $z=-4.89$ | 2 IN CTR $=150.00$ | NEXT $z=-6.72$ |  |  |  |  |  |
| *CHK Z JOB | 4 | IN | QUE | 9 | CLOCK | 21000 | D-C-10999 | $z=-5.16$ | 2 IN CTR= -7.10 | NEXT $Z=-6.00$ |  |  |  |  |  |
| *CHK Z JOB | 42 | IN | QUE | 2 | CLOCK | 21000 | D-C-10999 | $z=-12.50$ | 2 IN CTR $=-20.91$ | NEXT Z= -20.86 |  |  |  |  |  |
| *CHK 2 JOB | 46 | IN | QUE | 2 | CLOCK | 21000 | D-C-10999 | $z=-12.10$ | 2 IN CTR=-20.91 | NEXT Z $=-20.86$ |  |  |  |  |  |
| *CHK 2 JOB | 42 | IN Q | QUE | 2 | CLOCK | 21189 | D-C-11188 | $z=-12.50$ | 2 IN CTR=-21.84 | NEXT Z= -21.84 |  |  |  |  |  |
| *CHK 2 JOB | 46 | IN Q | QUE | 2 | CLOCK | 21189 | D-C-11188 | $z=-12.10$ | 2 IN CTR $=-21.84$ | NEXT Z $=-21.84$ |  |  |  |  |  |
| *CHK 2 JOB | 4 | IN | QUE | 9 | CLOCK | 21801 | D-C-11800 | $z=-5.16$ | 2 IN CTR $=150.00$ | NEXT $Z=-6.57$ |  |  |  |  |  |
| *CHK Z JOB | 4 | IN | QUE | 9 | CLOCK | 22000 | D-C-11999 | $z=-5.44$ | $Z$ IN CTR= -7.15 | NEXT $Z=-5.16$ |  |  |  |  |  |
| *CHK 2 JOB | 42 | IN | QUE | 2 | CLOCK | 22000 | D-C-11999 | $z=-13.35$ | 2 IN CTR=-21.84 | NEXT $Z=-13.89$ |  |  |  |  |  |
| *CHK 2 JOB | 46 | IN | QUE | 2 | CLOCK | 22000 | D-C-11999 | $z=-12.76$ | 2 IN CTR=-21.84 | NEXT Z= -13.89 |  |  |  |  |  |
| *CHK 2 JOB | 42 | IN | QUE | 2 | CLOCK | 22222 | D-C-12221 | $z=-13.35$ | 2 IN CTR=-14.87 | NEXT Z= -14.87 |  |  |  |  |  |
| *CHK Z JOB | 46 | IN | QUE | 2 | CLOCK | 22222 | D-C-12221 | $z=-12.76$ | 2 IN CTR=-14.87 | NEXT Z= -14.87 |  |  |  |  |  |
| *SETUP JOB | 4 | IN C | CTR | 9. | CLOCK | 22869 | D-C-12868 | $z=-5.44$ | S MEAN 500 | SPREAD 50 |  |  |  |  |  |
| *CHK 2 JOB | 42 | IN | QUE | 2 | CLOCK | 23000 | D-C-12999 | $z=-14.21$ | 2 IN CTR=-14.87 | NEXT $Z=-13.35$ |  |  |  |  |  |
| *CHK 2 JOB | 46 | IN | QUE | 2 | CLOCK | 23000 | D-C-12999 | $z=-13.42$ | 2 IN CTR=-14.87 | NEXT $Z=-13.35$ |  |  |  |  |  |
| *SETUP JOB | 42 | IN C | CTR | 2 | CLOCK | 23213 | D-C-13212 | $z=-14.21$ | S MEAN 500 | SPREAD 50 |  |  |  |  |  |
| *CHK 2 JOB | 46 | IN | QUE | 2 | CLOCK | 23213 | D-C-13212 | $Z=-13.42$ | 2 IN CTR=-14.21 | NEXT $2=-14.20$ |  |  |  |  |  |
| *PROC. JOB | 4 | IN C | CTR | 9 | CLOCK | 23370 | D-C-13369 | $z=-5.71$ | $P$ MEAN 100 | SPREAD 10 |  |  |  |  |  |
| *SEND JOB | 4 | TO Q | QUE | 5 | CLOCK | 23474 | D-C-13473 | $z=-5.71$ | T MEAN 2504.47 | ST.DEV 2259.38 | 0 | MEAN | . 00 | ST.DEV | . 00 |
| *CHK 2 JOB | 4 | IN Q | QUE | 5 | CLOCK | 23474 | D-C-13473 | $z=-5.71$ | 2 IN CTR=150.00 | NEXT Z= 150.00 |  |  |  |  |  |
| *SETUP JOB | 4 | IN C | CTR | 5 | CLOCK | 23474 | D-C-13473 | $z=-5.71$ | S MEAN 500 | SPREAD 50 |  |  |  |  |  |

## APPENDIX B

THE PERIODIC STATUS REPORT

The periodic status report may be as often as once at the end of each day or suppressed entirely. It is in two parts.

Part 1 contains 14 columns of information about the activities at centers in the system. Second entries are cumulative statistics.

Column Information

Center Number Utilization, Block P

Jobs in
Jobs out
Mean flow time
Flow time standard deviation
Mean queue time
Queue time standard deviation Number of jobs in queue Urgency number of next in line Identification of job in block $P$ Due date of current job

Urgency number of current job
Centers remaining for current job

Part 2 contains 9 columns of information about the job passing through the system without regard for the machine centers involved. Second entries are again cumulative statistics.

| Column | $\frac{\text { Information }}{1}$ |
| :---: | :--- |
| 2 | Job Size |
| 3 | Jobs in |
| 4 | Jobs out |
| 5 | Mean flow time |
| 6 | Flow time standard deviation |
| 7 | Mean D-A |
| 8 | Inter-exit mean |
| 9 | Inter-exit standard deviation |

****** REPORT FOR LAST 5 DAYS.


## APPENDIX C

## STATISTICAL TABLES

These tables are produced at the end of each simulation. There are two types of reports. The first type contains two tables. The first of these displays terminal data on the facilities in the system. Facilities 1 through 10 are blocks $S+P$. Facilities 21 through 30 are blocks P only. The second table contains terminal information about the queues in the system. The data in this table is not used in this study. Examination of the TOTAL ENTRIES column will show more entries per column in each center than passed through the system. This is because each time $z_{i}$ is computed and jobs re-sequenced in queues this action is taken as a new entry into the queue.

The second type constitute the primary source of data for this report. Examples of each variable on which terminal statistics are tabulated are shown in numerical order.

The following tables are used:

| Table Number |  |
| ---: | :--- |
| $1-10$ |  |
| $27-30$ | Center flow time |
| $41-50$ | Center idle time |
| $67-63$ | Total flow time by job size |


| $66-68$ | D-A by job size |
| :--- | :--- |
| $71-73$ | Total flow time by initial z |
| $76-78$ | D-A by initial z |
| 81 | Total flow time |
| 82 | D-A |
| 83 | Inter-exit time |


| $\begin{array}{r} \text { FACILITY } \\ \text { NR } \end{array}$ | AVERAGE UTILIZATION |  | NUMBER ENTRIES | AVERAGE TIME/TRANS | TRANS | \$TRANS |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | . 9112 |  | 820 | 1000.09 | 72,5 | 0 |  |  |
| 2 | . 9252 |  | 835 | 997.23 | 18.5 | 0 |  |  |
| 3 | . 9350 |  | 844 | 997.05 | 59.5 | 0 |  |  |
| 4 | . 8861 |  | 800 | 996.84 | 44.5 | 0 |  |  |
| 5 | - 9043 |  | 815 | 998.61 | 61.5 | 0 |  |  |
| 6 | -9069 |  | 817 | 998.98 | 0 | 0 |  |  |
| 7 | -9289 |  | 836 | 1000.00 | 25.5 | 0 |  |  |
| 8 | . 9089 |  | 820 | 997.56 | 39.5 | 0 |  |  |
| 9 | . 9716 |  | 877 | 997.07 | 6.5 | 0 |  |  |
| 10 | . 8999 |  | 811 | 998.69 | $73 \cdot 5$ | 0 |  |  |
| 21 | . 4552 |  | 820 | 499.56 | 72,H | 0 |  |  |
| 22 | . 4619 |  | 834 | 498.47 | 0 | 0 |  |  |
| 23 | . 4667 |  | 843 | 498.28 | 0 | 0 |  |  |
| 24 | . 4427 |  | 799 | 498.68 | 0 | 0 |  |  |
| 25 | . 4506 |  | 814 | 498.19 | 0 | 0 |  |  |
| 26 | . 4533 |  | 817 | 499.38 | 0 | 0 |  |  |
| 27 | . 4636 |  | 836 | 499.08 | 25.H | 0 |  |  |
| 28 | . 4544 |  | 819 | 499.36 | 0 | 0 |  |  |
| 29 | . 4832 |  | 877 | 495.87 | 6.H | 0 |  |  |
| 30 | . 4493 |  | 810 | 499.23 | 0 | 0 |  |  |
| 41 | . 0000 |  | 1568 | . 00 | 0 | 0 |  |  |
| QUEUE | MAXIMUM $A$ <br> CONTENTS | AVERAGE CONTENTS | TOTAL ENTRIES | $\begin{gathered} \text { ZERO } \\ \text { ENTRIES } \end{gathered}$ | PERCENT <br> ZEROS | AVERAGE TIME/TRANS | TABLE NUMBER | CURRENT CONTENTS |
| NR | CONTENTS CO | ONTENTS | ENTRIES | ENTRIES | ZEROS | TIME/TRANS | NUMBER | CONTENTS |
| 1 | 6 | $.75$ | 3024 | 83 | 2.7 | 223.53 | - 0 | $1$ |
| 2 | 6 | . 83 | - 3317 | 60 | 1.8 | 224.68 | 0 | 1 |
| 3 | 6 | . 83 | 3406 | 59 | 1.7 | 219.08 | 0 | 1 |
| 4 | 6 | . 74 | 2876 | 84 | 2.9 | 230.06 | 0 | 1 |
| 5 | 6 | . 78 | 3095 | 65 | 2.1 | 226.27 | 0 | 1 |
| 6 | 6 | . 75 | 2927 | 84 | $2 \cdot 9$ | 231.23 | 0 | 0 |
| 7 | 6 | . 82 | - 3130 | 63 | 2.0 | 235.26 | 0 | 1 |
| 8 | 5 | . 73 | 2739 | 88 | 3.2 | 240.63 | 0 | 1 |
| 9 | 7 | . 93 | 3993 | 22 | - 6 | 208.80 | 0 | 1 |
| 10 | 5 | . 75 | 2855 | 82 | 2.9 | 236.68 | 0 | 1 |
| 21 | 14 | 2.88 | 5453 | 0 | . 0 | 475.68 | 0 | 1 |
| 22 | 16 | 3.74 | 7072 | 0 | . 0 | 475.58 | 0 | 1 |
| 23 | 15 | 3.65 | 6933 | 1 | . 0 | 473.77 | 0 | 5 |
| 24 | 14 | 2.95 | 5617 | 0 | - 0 | 472.77 | 0 | 8 |
| 25 | 15 | 2.83 | 5373 | 1 | . 0 | 473.50 | 0 | 0 |
| 26 | 16 | 3.05 | 5731 | 1 | - 0 | 478.90 | 0 | 0 |
| 27 | 9 | 2.67 | 5133 | 1 | - 0 | 467.65 | 0 | 2 |
| 28 | 13 | 2.16 | 4166 | 0 | . 0 | 465.71 | 0 | 3 |
| 29 | 23 | 7.19 | 13345 | 2 | - 0 | 485.19 | 0 | 7 |
| 30 | 17 | 3.08 | 5866 | 0 | . 0 | 472.86 | 0 | 5 |
| 41 | 61 | . 00 | 80031 | 8320 | 10.4 | - 00 | 0 | 0 |

TABLE NUMBER

| ENTRIES IN | $\begin{array}{r} \text { TABLE } \\ 825 \end{array}$ | MEAN ARGUMENT4385.028 |  | $\begin{array}{r} \text { STANDARD DEVIATION } \\ 4046.056 \end{array}$ |  | NON-WEIGHTED |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 420300 | 4303.644 |  | 132100.950 |  | WEI GHTED |  |
|  | UPPER | OBSERVED | PERCENT | CUMULATIVE | CuMuLative | MULT IPLE | DEVIATION |
|  | LIMIT | FREQUENCY | OF TOTAL | PERCENTAGE | REMAINDER | OF MEAN | FROM MEAN |
|  | 1000 | 19000 | 4.52 | 4.5 | 95.5 | . 228 | -. 837 |
|  | 2000 | 171600 | 40.83 | 45.3 | 54.7 | . 456 | -. 589 |
|  | 3000 | 46500 | 11.06 | 56.4 | 43.6 | . 684 | -. 342 |
|  | 4000 | 34000 | 8.09 | 64.5 | 35.5 | . 912 | . 095 |
|  | 5000 | 22000 | 5.23 | 69.7 | 30.3 | 1.140 | . 152 |
|  | 6000 | 18200 | 4.33 | 74.1 | 25.9 | 1.368 | . 399 |
|  | 7000 | 17500 | 4.16 | 78.2 | 21.8 | 1.596 | . 646 |
|  | 8000 | 15500 | 3.69 | 81.9 | 18.1 | 1.824 | . 893 |
|  | 9000 | 12500 | 2.97 | 84.9 | 15.1 | 2.052 | 1.141 |
|  | 10000 | 12000 | 2.86 | 87.7 | 12.3 | 2.280 | 1.388 |
|  | 11000 | 7000 | 1.67 | 89.4 | 10.6 | 2.50 .9 | 1.635 |
|  | 12000 | 8000 | 1.90 | $91 \cdot 3$ | 8.7 | 2.737 | 1.882 |
|  | 13000 | 10500 | $2 \cdot 50$ |  | 6.2 | 2.965 | 2.129 |
|  | 14000 | 9500 | 2.26 | 96.1 | 3.9 | 3.193 | 2.376 |
|  | 15000 | 7500 | 1.78 | 97.9 | 2.1 | 3.421 | 2.624 |
|  | 16000 | 4500 | 1.07 | 98.9 | 1.1 | 3.649 | 2.871 |
|  | 17000 | 4500 | 1.07 | 100.0 | . 0 | 3.877 | 3.118 |
| REMAINING | FREQUENCIES | ARE ALL 2 | ZERO |  |  |  |  |
| TABLE NUMBER |  |  |  |  |  |  |  |
| ENTRIES IN | TABLE | MEAN | $\begin{aligned} & \text { ARGUMENT } \\ & 3821.123 \end{aligned}$ | $\begin{array}{r} \text { STANPARD DEVIATION } \\ 3632.483 \end{array}$ |  | NON-WEIGHTED |  |
|  | 405400 |  | 3762.668 | 116959.225 |  | WEIGHTED |  |
|  | UPPER <br> LIMIT | 9BSERVED FREQUENCY | PERCENT OF TOTAL | GUMULATIVE PERGENTAGE | CUMULATIVE REMAINDER | MULTIPLE OF MEAN | DEVIATION <br> FROM MEAN |
|  | 1000 | 24700 | $6.09$ | 6.1 | 93.9 | $.262$ | -. 777 |
|  | 2000 | 181600 | 44.80 | 50.9 | 49.1 | . 523 | -. 501 |
|  | 3090 | 52100 | 12.85 | 63.7 | 36.3 | .785 | -. 226 |
|  | 400 | 26000 | 6.41 | 70.2 | 29.8 | 1.047 | . 049 |
|  | 5090 | 19000 | 4.69 | 74.8 | $25 \cdot 2$ | 1.309 | . 325 |
|  | 6000 | 15500 | 3.82 | 78.7 | $21 \cdot 3$ | 1.570 | . 600 |
|  | 700 | 16000 | 3.95 | 8 8.6 | $17.4$ | 1.832 | . 875 |
|  | 8000 | 11000 | 2.71 | 85.3 | 14.7 | 2.094 | 1.150 |
|  | 9000 | 9000 | 2.22 | 87.5 | 12.5 | 2.355 | 1.426 |
|  | 19000 | 10000 | 2.47 | 90.0 | 10.0 7.5 | 2.617 2.879 | 1.701 1.976 |
|  | 11000 | 10000 8500 | 2.47 2.10 | 92.5 9.4 .6 | 7.5 5.4 | 2.879 3.140 | 1.976 2.252 |
|  | 12000 13000 | 8500 7500 | 2.10 1.85 | 94.6 96.4 | 5.4 3.6 | 3.140 3.402 | 2. 2.527 |
|  | 14000 15000 | 5500 | 1.36 | 97.8 | 2.2 | 3.664 | 2.802 |
|  | 15000 | 5500 | 1.36 | 99.1 | -9 | 3.926 | 3.077 |
|  | 16000 | 2000 | . 4.49 | 99.6 | - 4 | 4.187 | 3.353 |
|  | 17000 | 1500 | . 37 | 100.0 | . 0 | 4.449 | 3.628 |
| REMAINING | FREQUENCIES | ARE ALL 2 | ZERO |  |  |  |  |


| ENTRIES IN | $\begin{array}{r} V \text { TABLE } \\ 802 \end{array}$ | MEAN | $\begin{array}{r} \text { ARGUMENT } \\ 596.868 \end{array}$ | STANDARD DEV |  | NON-WEIGHTED |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | UPPER | OBSERVED | PERCENT | CUMULATIVE | CUMULATIVE | MULT IPLE | DEVIATION |
|  | LIMIT | FREQUENCY | OF TOTAL | PERCENTAGE | REMAINDER | OF MEAN | FROM MEAN |
|  | 100 | 0 | . 00 | . 0 | 100.0 | . 168 | -1.253 |
|  | 200 | 0 | . 00 | - 0 | 100.0 | - 335 | -1.001 |
|  | 300 | 0 | . 00 | . 0 | 100.0 | . 503 | -. 749 |
|  | 400 | 0 | . 00 | . 0 | 100.0 | . 670 | -. 496 |
|  | 500 | 362 | 44.15 | 44.1 | 55.9 | . 838 | -. 244 |
|  | 600 | 390 | 47.56 | 91.7 | 8.3 | 1.005 | . 008 |
|  | 700 | 4 | . 49 | 92.2 | $7 \cdot 8$ | 1.173 | . 260 |
|  | 800 | 6 | .73 | 92.9 | $7 \cdot 1$ | 1.340 | . 512 |
|  | 900 | 3 | . 37 | 93.3 | $6 \cdot 7$ | 1.508 | . 764 |
|  | 1000 | 4 | . 49 | 93.8 | 6.2 | 1.675 | 1.017 |
|  | 1100 | 4 | . 49 | 94.3 | 5.7 | $=1.843$ | 1.269 |
|  | 1200 | 5 | . 61 | 94.9 | 5.1 | 2.010 | 1.521 |
|  | 1300 | 2 | . 24 | 95.1 | 4.9 | 2.178 | 1.773 |
|  | 1400 | 5 | . 61 | 95.7 | $4 \cdot 3$ | 2.346 | 2.025 |
|  | 1500 | 7 | - 85 | 96.6 | 3.4 | 2.513 | 2.277 |
|  | 1600 | 1 | . 12 | 96.7 | $3 \cdot 3$ | 2.681 | 2.529 |
|  | 1700 | 1 | -12 | 96.8 | 3.2 | 2.848 | 2.782 |
|  | 1800 | 1 | . 12 | 97.0 | 3.0 | 3.016 | 3.034 |
|  | 1900 | 2 | . 24 | 97.2 | $2 \cdot 8$ | 3.183 | 3.286 |
|  | 2000 | 1 | - 12 | 97.3 | $2 \cdot 7$ | 3.351 | 3.538 |
|  | 2100 | 5 | . 61 | 97.9 | $2 \cdot 1$ | 3.518 | 3.790 |
|  | 2200 | 0 | . 00 | 97.9 | $2 \cdot 1$ | 3.686 | 4.042 |
|  | 2300 | 0 | . 00 | 97.9 | 2.1 | 3.853 | 4.295 |
|  | 2400 | 1 | . 12 | 98.0 | 2.0 | 4.021 | 4.547 |
|  | 2500 | 1 | - 12 | 98.2 | 1.8 | 4.189 | 4.799 |
|  | 2600 | 3 | . 37 | 98.5 | 1.5 | 4.356 | 5.051 |
|  | 2700 | 1 | . 12 | 98.7 | 1.3 | 4.524 | 5.303 |
|  | 2800 | 1 | . 12 | 98.8 | 1.2 | 4.691 | 5.555 |
|  | 2900 | 1 | - 12 | 98.9 | 1.1 | 4.859 | 5.808 |
|  | 3000 | 2 | . 24 | 99.1 | . 9 | 5.026 | 6.060 |
|  | 3100 | 1 | - 12 | 99.3 | . 7 | 5.194 | 6.312 |
|  | 3200 | 2 | . 24 | 99.5 | - 5 | 5.361 | 6.564 |
|  | 3300 | 1 | - 12 | 99.6 | . 4 | 5.529 | 6.816 |
|  | 3400 | 1 | . 12 | 99.8 | . 2 | 5.696 | 7.068 |
|  | 3500 | 2 | . 24 | 100.0 | - 0 | 5.864 | 7.320 |
| REMAINING | FREQUEN | ARE ALL | ZERO |  |  |  |  |

TABLE NUMBER 42

| ENTRIES | $\begin{array}{r} \text { IN TABLE } \\ 817 \end{array}$ | MEAN | ARGUMENT $3523.820$ |
| :---: | :---: | :---: | :---: |
|  | 415900 |  | 3454.700 |
|  | UPPER | OBSERVED | PERCENT |
|  | LIMIT | FREQUENCY | OF TOTAL |
|  | 0 | 29500 | 7.09 |
|  | 500 | 74600 | 17.94 |
|  | 1000 | 72100 | 17.34 |
|  | 1500 | 30000 | 7.21 |
|  | 2000 | 24000 | 5.77 |
|  | 2500 | 11000 | 2.64 |
|  | 3000 | 14500 | 3.49 |
|  | 3500 | 12100 | 2.91 |
|  | 4000 | 13600 | 3.27 |
|  | 4500 | 11500 | 2.77 |
|  | 5000 | 12000 | 2.89 |
|  | 5500 | 9500 | 2.28 |
|  | 6000 | 8500 | 2.04 |
|  | 6500 | 8000 | 1.92 |
|  | 7000 | 7000 | 1.68 |
|  | 7500 | 11000 | 2.64 |
|  | 8000 | 6500 | 1.56 |
|  | 8500 | 4000 | . 96 |
|  | 9000 | 5000 | 1.20 |
|  | 9500 | 3000 | . 72 |
|  | 10000 | 5500 | 1.32 |
|  | 10500 | 5000 | 1.20 |
|  | 11000 | 5000 | 1.20 |
|  | 11500 | 2000 | . 48 |
|  | 12000 | 5000 | 1.20 |
|  | 12500 | 5000 | 1.20 |
|  | 13000 | 2000 | . 48 |
|  | 13500 | 3000 | . 72 |
|  | 14000 | 3500 | . 84 |
|  | 14500 | 3500 | . 84 |
|  | 15000 | 2500 | . 60 |
|  | 15500 | 2500 | . 60 |
|  | 16000 | 2500 | . 60 |
|  | 16500 | 1500 | . 36 |

REMAINING FREQUENCIES ARE ALL ZERO

| STANDARD DEVIATION4074.316 |  | NON-WEIGHTED |  |
| :---: | :---: | :---: | :---: |
| 119 |  | WEIGHTED |  |
| CUMULATIVE | CUMULATIVE | MULTIPLE | DEVIATION |
| PERCENTAGE | REMAINDER | OF MEAN | FROM MEAN |
| 7.1 | 92.9 | . 000 | -. 865 |
| 25.0 | 75.0 | . 142 | -. 742 |
| 42.4 | 57.6 | . 284 | -. 619 |
| 49.6 | 50.4 | . 426 | -. 497 |
| 55.3 | 44.7 | . 568 | -. 374 |
| 58.0 | 42.0 | . 709 | -. 251 |
| 61.5 | 38.5 | - 851 | -. 129 |
| 64.4 | 35.6 | . 993 | -. 006 |
| 67.7 | 32.3 | 1.135 | . 117 |
| 70.4 | 29.6 | 1.277 | . 240 |
| 73.3 | 26.7 | 1.419 | . 362 |
| 75.6 | 24.4 | 1.561 | . 485 |
| 77.6 | 22.4 | 1.703 | . 608 |
| 79.6 | 20.4 | 1.845 | . 730 |
| 81.2 | 18.8 | 1.986 | . 853 |
| 83.9 | 16.1 | 2.128 | . 976 |
| 85.5 | 14.5 | 2.270 | 1.099 |
| 86.4 | 13.6 | 2.412 | 1.221 |
| 87.6 | 12.4 | 2.554 | 1.344 |
| 88.3 | 11.7 | 2.696 | 1.467 |
| 89.7 | 10.3 | 2.838 | 1.590 |
| 90.9 | $9 \cdot 1$ | 2.980 | 1.712 |
| 92.1 | $7 \cdot 9$ | 3.122 | 1.835 |
| 92.5 | 7.5 | 3.264 | 1.958 |
| 93.7 | 6.3 | 3.405 | 2.080 |
| 95.0 | 5.0 | 3.547 | 2.203 |
| 95.4 | 4.6 | 3.689 | 2.326 |
| 96.2 | 3.8 | 3.831 | 2.449 |
| 97.0 | 3.0 | 3.973 | 2.571 |
| 97.8 | 2.2 | 4.115 | 2.694 |
| 98.4 | 1.6 | 4.257 | 2.817 |
| 99.0 | 1.0 | 4.399 | 2.939 |
| 99.6 | . 4 | 4.541 | 3.062 |
| 100.0 | . 0 | 4.682 | 3.185 |

TABLE NUMBER 6

| ENTRIES | IN TABLE 1484 | MEAN | $\begin{aligned} & \text { IENT } \\ & 296 \end{aligned}$ |
| :---: | :---: | :---: | :---: |
|  | 756500 |  | 591 |
|  | UPPER | OBSERVED | PERCENT |
|  | LIMIT | FREQUENCY | OF TOTAL |
|  | 1000 | 3000 | . 40 |
|  | 2000 | 21000 | 2.78 |
|  | 3000 | 12000 | 1.59 |
|  | 4000 | 9000 | 1.19 |
|  | 5000 | 15000 | 1.98 |
|  | 6000 | 11500 | 1.52 |
|  | 7000 | 10000 | 1.32 |
|  | 8000 | 10000 | 1.32 |
|  | 9000 | 10000 | 1.32 |
|  | 10000 | 7500 | - 99 |
|  | 11000 | 14500 | 1.92 |
|  | 12000 | 12000 | 1.59 |
|  | 13000 | 15000 | 1.98 |
|  | 14000 | 11500 | 1.52 |
|  | 15000 | 15500 | 2.05 |
|  | 16000 | 8500 | 1.12 |
|  | 17000 | 14500 | 1.92 |
|  | 18000 | 15000 | 1.98 |
|  | 19000 | 10500 | 1.39 |
|  | 20000 | 16000 | 2.12 |
|  | 21000 | 13500 | 1.78 |
|  | 22000 | 20000 | 2.64 |
|  | 23000 | 18500 | 2.45 |
|  | 24000 | 12500 | 1.65 |
|  | 25000 | 12000 | 1.59 |
|  | 26000 | 16000 | 2.12 |
|  | 27000 | 14500 | 1.92 |
|  | 28000 | 12500 | 1.65 |
|  | 29000 | 12000 | 1.59 |
|  | 30000 | 14000 | 1.85 |
|  | 31000 | 16000 | 2.12 |
|  | 32000 | 11500 | 1.52 |
|  | 33000 | 13000 | 1.72 |
|  | 34000 | 13000 | 1.72 |
|  | 35000 | 14000 | 1.85 |
|  | 36000 | 12000 | 1.59 |
|  | 37000 | 17500 | 2.31 |
|  | 38000 | 13000 | 1.72 |
|  | 39000 | 15000 | 1.98 |
|  | 40000 | 10000 | 1.32 |
|  | OVERFLOW | 234000 | 30.93 |

ST
TANDARD DEVIATION
16668.744
NON-WEI GHTED

| 762051.400 |  | WEI GHTED |  |
| :---: | :---: | :---: | :---: |
| CUMULATIVE | CUMULATIVE | MULTIPLE | DEVIATION |
| PERCENTAGE | REMAINDER | OF MEAN | FROM MEAN |
| . 4 | 99.6 | . 033 | -1.748 |
| 3.2 | 96.8 | . 066 | -1.688 |
| 4.8 | 95.2 | . 100 | -1.628 |
| 5.9 | 94.1 | .133 | -1.568 |
| 7.9 | 92.1 | . 166 | -1.508 |
| 9.5 | 90.5 | . 199 | -1.448 |
| 10.8 | 89.2 | . 232 | -1.388 |
| 12.1 | 87.9 | . 265 | -1.328 |
| 13.4 | 86.6 | . 299 | -1.268 |
| 14.4 | 85.6 | . 332 | -1.208 |
| 16.3 | 83.7 | . 365 | -1.148 |
| 17.9 | 82.1 | . 398 | -1.088 |
| 19.9 | 80.1 | . 431 | -1.028 |
| 21.4 | 78.6 | . 465 | -. 968 |
| 23.5 | 76.5 | . 498 | -. 908 |
| 24.6 | 75.4 | . 531 | -. 848 |
| 26.5 | 73.5 | . 564 | -. 788 |
| 28.5 | 71.5 | . 597 | -. 728 |
| 29.9 | 70.1 | . 631 | -. 668 |
| 32.0 | 68.0 | . 664 | -. 608 |
| 33.8 | 66.2 | . 697 | -. 548 |
| 36.4 | 63.6 | . 730 | -. 488 |
| 38.9 | 61.1 | . 763 | -. 428 |
| 40.5 | 59.5 | . 796 | -. 368 |
| 42.1 | 57.9 | . 830 | -. 308 |
| 44.2 | 55.8 | . 863 | -. 248 |
| 46.1 | 53.9 | . 896 | -. 188 |
| 47.8 | 52.2 | . 929 | -. 128 |
| 49.4 | 50.6 | . 962 | -. 068 |
| 51.2 | 48.8 | . 996 | -. 008 |
| 53.3 | 46.7 | 1.029 | . 052 |
| 54.9 | 45.1 | 1.062 | . 112 |
| 56.6 | 43.4 | 1.095 | . 172 |
| 58.3 | 41.7 | 1.128 | . 232 |
| 60.1 | 39.9 | 1.161 | . 292 |
| 61.7 | 38.3 | 1.195 | . 352 |
| 64.0 | 36.0 | 1.228 | . 412 |
| 65.8 | 34.2 | 1.261 | . 472 |
| 67.7 | 32.3 | 1.294 | . 532 |
| 69.1 | 30.9 | 1.327 | - 592 |
| 100.0 |  |  |  |

TABLE NUMBER 63

ENTRIES IN TABLE 4

| 500 | 13544.400 |  |
| ---: | ---: | ---: |
| UPPER | OBSERVED | PERCENT |
| LIMIT | FREQUENCY | OF TOTAL |
| 1000 | 0 | .00 |
| 2000 | 100 | 20.00 |
| 3000 | 0 | .00 |
| 4000 | 0 | .00 |
| 5000 | 0 | .00 |
| 6000 | 0 | .00 |
| 7000 | 0 | .00 |
| 8000 | 0 | .00 |
| 9000 | 0 | .00 |
| 10000 | 0 | .00 |
| 11000 | 100 | 20.00 |
| 12000 | 0 | .00 |
| 13000 | 0 | .00 |
| 14000 | 0 | .00 |
| 15000 | 0 | .00 |
| 16000 | 100 | 20.00 |
| 17000 | 0 | 20.00 |
| 18000 | 0 | .00 |
| 19000 | 0 | .00 |
| 20000 | 0 | .00 |
| 21000 | 0 | .00 |
| 22000 | 0 | .00 |
| 23000 | 0 | .00 |
| 24000 | 100 | .00 |
| 25000 | 0 | 20.00 |

STANDARD DEVIATION
4897.578
157056.670

| CUMULATIVE | CUMULATIVE |
| ---: | ---: |
| PERCENTAGE | REMAINDER |
| 20.0 | 100.0 |
| 20.0 | 80.0 |
| 20.0 | 80.0 |
| 20.0 | 80.0 |
| 20.0 | 80.0 |
| 20.0 | 80.0 |
| 20.0 | 80.0 |
| 20.0 | 80.0 |
| 20.0 | 80.0 |
| 40.0 | 80.0 |
| 40.0 | 60.0 |
| 40.0 | 60.0 |
| 40.0 | 60.0 |
| 40.0 | 60.0 |
| 60.0 | 60.0 |
| 80.0 | 40.0 |
| 80.0 | 20.0 |
| 80.0 | 20.0 |
| 80.0 | 20.0 |
| 80.0 | 20.0 |
| 80.0 | 20.0 |
| 80.0 | 20.0 |
| 80.0 | 20.0 |
| 100.0 | 20.0 |
|  | -.0 |

## NON-WEIGHTED

WEIGHTED

| MULTIPLE | DEVIATION |
| ---: | ---: |
| OF MEAN | FROM MEAN |
| .059 | -3.253 |
| .118 | -3.049 |
| .177 | -2.844 |
| .236 | -2.640 |
| .295 | -2.436 |
| .354 | -2.232 |
| .413 | -2.028 |
| .473 | -1.823 |
| .532 | -1.619 |
| .591 | -1.415 |
| .650 | -1.211 |
| .709 | -1.007 |
| .768 | -.803 |
| .827 | -.598 |
| .886 | -.394 |
| 1.945 | -.190 |
| 1.004 | .014 |
| 1.063 | .218 |
| 1.122 | .423 |
| 1.181 | .627 |
| 1.240 | .831 |
| 1.299 | 1.035 |
| 1.358 | 1.239 |
| 1.418 | 1.443 |
| 1.477 | 1.648 |

TABLE NUMBER 66

| ENTRIES | $\begin{array}{r} \text { IN TABLE } \\ 1484 \end{array}$ | MEAN ARGUMENT$-3920.720$ |  | STANDARD DEVIATION6399.737 |  | WE I GHTED |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 756500 | -3845.564 |  | 166161.650 |  | WE I GHTED |  |
|  | UPPER | OBSERVED | PERCENT | CUMULATIVE | CUMULATIVE | MULTIPLE | DEVIATION |
|  | LIMIT | FREQUENCY | OF TOTAL | PERCENTAGE | REMAINDER | OF MEAN | FROM MEAN |
|  | -20000 | 17500 | 2.31 | 2.3 | 97.7 | 5.101 | -2.512 |
|  | -19000 | 4000 | . 53 | 2.8 | 97.2 | 4.846 | -2.356 |
|  | -18000 | 4000 | . 53 | 3.4 | 96.6 | 4.591 | -2.200 |
|  | -17000 | 9000 | 1.19 | 4.6 | 95.4 | 4.336 | -2.044 |
|  | -16000 | 8500 | 1.12 | 5.7 | 94.3 | 4.081 | -1.887 |
|  | -15000 | 9000 | 1.19 | 6.9 | 93.1 | 3.826 | -1.731 |
|  | -14000 | 15000 | 1.98 | 8.9 | 91.1 | 3.571 | -1.575 |
|  | -13000 | 13000 | 1.72 | 10.6 | 89.4 | 3.316 | -1.419 |
|  | -12000 | 11000 | 1.45 | 12.0 | 88.0 | 3.061 | -1.262 |
|  | -11000 | 23000 | 3.04 | 15.1 | 84.9 | 2.806 | -1.106 |
|  | -10000 | 22000 | 2.91 | 18.0 | 82.0 | 2.551 | -. 950 |
|  | -9000 | 20000 | 2.64 | 20.6 | 79.4 | 2.295 | -. 794 |
|  | -8000 | 23000 | 3.04 | 23.7 | 76.3 | 2.040 | -. 637 |
|  | -7000 | 26000 | 3.44 | 27.1 | 72.9 | 1.785 | -. 481 |
|  | -6000 | 17500 | 2.31 | 29.4 | 70.6 | 1.530 | -. 325 |
|  | -5000 | 36000 | 4.76 | 34.2 | 65.8 | 1.275 | -. 169 |
|  | -4000 | 46500 | 6.15 | 40.3 | 59.7 | 1.020 | -. 012 |
|  | -3000 | 62500 | 8.26 | 48.6 | 51.4 | . 765 | . 144 |
|  | -2000 | 65500 | 8.66 | 57.2 | 42.8 | . 510 | . 300 |
|  | -1000 | 61500 | 8.13 | 65.4 | 34.6 | . 255 | . 456 |
|  | 0 | 61000 | 8.06 | 73.4 | 26.6 | -. 000 | . 613 |
|  | 1000 | 47000 | 6.21 | 79.6 | 20.4 | -. 255 | . 769 |
|  | 2000 | 38500 | 5.09 | 84.7 | 15.3 | -. 510 | . 925 |
|  | 3000 | 35000 | 4.63 | 89.4 | $10 \cdot 6$ | -. 765 | 1.081 |
|  | 4000 | 33500 | 4.43 | 93.8 | 6.2 | -1.020 | 1.238 |
|  | 5000 | 20500 | 2.71 | 96.5 | 3.5 | -1.275 | 1.394 |
|  | 6000 | 8500 | 1.12 | 97.6 | 2.4 | -1.530 | 1.550 |
|  | 7000 | 6500 | - 86 | 98.5 | 1.5 | -1.785 | 1.706 |
|  | 8000 | 4000 | - 53 | 99.0 | 1.0 | -2.040 | 1.863 |
|  | 9000 | 4000 | . 53 | 99.5 | - 5 | -2.295 | 2.019 |
|  | 10000 | 1000 | . 13 | 99.7 | - 3 | -2.551 | 2.175 |
|  | 11000 | 500 | - 07 | 99.7 | - 3 | -2.806 | 2.331 |
|  | 12000 | 500 | . 07 | 99.8 | - 2 | -3.061 | 2.488 |
|  | 13000 | 500 | . 07 | 99.9 | - 1 | -3.316 | 2.644 |
|  | 14000 | 0 | . 00 | 99.9 | - 1 | -3.571 | 2.800 |
|  | 15000 | 500 | . 07 | 99.9 | -1 | -3.826 | 2.956 |
|  | 16000 | 0 | . 00 | 99.9 | -1 | -4.081 | 3.113 |
|  | 17000 | 0 | . 00 | 99.9 | -1 | -4.336 | 3.269 |
|  | 18000 | 500 | - 07 | 100.0 | - 0 | -4.591 | 3.425 |

table number 68

| ENTRIES IN TABLE 4 | MEAN ARGUMENT-14430.250 |  |
| :---: | :---: | :---: |
| 500 |  | 200 |
| UPPER | observed | PERCENT |
| LIMIT | FREQUENCY | OF TOTAL |
| -20000 | 0 | . 00 |
| -19000 | 0 | . 00 |
| -18000 | 0 | . 00 |
| -17000 | 0 | . 00 |
| -16000 | 100 | 20.00 |
| -15000 | 100 | 20.00 |
| -14000 | 100 | 20.00 |
| -13000 | 0 | . 00 |
| -12000 | 0 | . 00 |
| -11000 | 0 | . 00 |
| -10000 | 100 | 20.00 |
| -9000 | 0 | . 00 |
| -8000 | - | . 00 |
| -7000 | 0 | . 00 |
| -6000 | 0 | . 00 |
| -5000 | 0 | . 00 |
| -4000 | - | . 00 |
| -3000 | 0 | . 00 |
| -2000 | 0 | . 00 |
| -1000 | 100 | 20.00 |

remaining frequencies are all zero

| STANDARD DEVIATION |  | NON-WEIGHTED |  |
| :---: | :---: | :---: | :---: |
| 1300 |  | WEIGHTED |  |
| Cumulative PERCENTAGE | CUMULATIVE REMAINDER | MULTIPLE OF MEAN | DEVIATION <br> FROM MEAN |
| . 0 | 100.0 | 1.386 | -2.502 |
| . 0 | 100.0 | 1.317 | -2.053 |
| . 0 | 100.0 | 1.247 | -1.604 |
| . 0 | 100.0 | 1.178 | -1.155 |
| 20.0 | 80.0 | 1.109 | -. 705 |
| 40.0 | 60.0 | 1.039 | -. 256 |
| 60.0 | 40.0 | . 970 | . 193 |
| 60.0 | 40.0 | . 901 | . 643 |
| 60.0 | 40.0 | . 832 | 1.092 |
| 60.0 | 40.0 | . 762 | 1.541 |
| 80.0 | 20.0 | . 693 | 1.990 |
| 80.0 | 20.0 | . 624 | 2.440 |
| 80.0 | 20.0 | . 554 | 2.889 |
| 80.0 | 20.0 | . 485 | 3.338 |
| 80.0 | 20.0 | . 416 | 3.788 |
| 80.0 | 20.0 | . 346 | 4.237 |
| 80.0 | 20.0 | - 277 | 4.686 |
| 80.0 | 20.0 | . 208 | 5.136 |
| 80.0 | 20.0 | . 139 | 5.585 |
| 100.0 | -. 0 | . 069 | 6.034 |

TABLE NUMBER 71


1488
757000
UPPER
LIMIT
1000
2000
3000
4000
5000
6000
7000
8000
9000
10000
11000
12000
13000
14000
15000
16000
17000
18000
19000
20000
21000
22000
23000
24000
25000
26000
27000
28000
29000
30000
31000
32000
33000
34000
35000
36000
37000
38000
39000
40000
OVERFLOW

MEAN ARGUMENT
30098.802
29546.014
PERCENT
OF TOTAL
.40
2.79
1.59
1.19
1.98
1.52
1.32
1.32
1.32
1.99
1.93
1.59
1.98
1.52
2.05
1.14
1.93
1.98
1.39
2.11
1.78
2.64
2.44
1.65
1.60
2.11
1.92
1.65
1.59
1.85
2.11
1.52
1.72
1.72
1.85
1.59
2.31
1.72
1.98
1.32
30

STANDARD DEVIATION
16662.293
761810.490

CUMULATIVE PERCENTAGE
.4
3.2
4.8
6.0
7.9
9.5
10.8
12.1
13.4
14.4
16.3
17.9
19.9
21.4
23.5
24.6
26.5
28.5
29.9
32.0
33.8
36.4
38.9
40.5
42.1
44.3
46.2
47.8
49.4
51.3
53.4
54.9
56.6
58.3
60.2
61.8
64.1
65.8
67.8
69.1
100.0

CUMULATIVE REMA INDER 99.6
96.8 95.2
94.0 $\begin{array}{ll} & .036 \\ 94.0 & .100 \\ 9.133\end{array}$ $92.1 \quad .13$

| 90.5 | .166 |
| :--- | :--- |
| 89.2 | .199 |

$89.2 \quad .23$
87.9
86.6
85.6
83.7
83.7
82.1
82.1
80.1
80.1
78.6
76.5
76.5
75.4
75.4
73.5
73.5
71.5
70.1
68.0

68
$\begin{array}{lll} & .664 & -.606 \\ 63.6 & .698 & -.546\end{array}$
$\begin{array}{lll}63.6 & .731 & -.486 \\ 61.1 & .764 & . .426\end{array}$
$61.1 \quad .764 \quad-.426$

| 59.5 | .764 | -.426 |
| :--- | :--- | :--- |
| 57.9 | -.797 | -.366 |


| 57.9 | .797 | -.36 |
| :--- | :--- | :--- |
|  | .831 | -.306 |


| 55.7 | .831 | -.306 |
| :--- | :--- | :--- |
| 53.8 | .864 | -.246 |


| 53.8 | .897 | -.186 |
| :--- | :--- | :--- |
| 52.2 | .930 | -.126 |

50.6 -930 -.126
$48.7 \quad .963 \quad=.066$
$46.7-.097$

| 1.030 | .054 |
| :--- | :--- |
| 1.063 | .114 |

1.063
1.096
1.130
1.163
$1.196 \quad .354$
$1.229 \quad .414$
1.263
$1.329 \quad \square$

TABLE NUMBER 8

| ENTRIES | $\begin{array}{r} \text { IN TABLE } \\ 1488 \end{array}$ | MEAN | $\begin{aligned} & \text { MENT } \\ & .802 \end{aligned}$ |
| :---: | :---: | :---: | :---: |
|  | 757000 |  | 014 |
|  | UPPER | ObSERVED | PERCENT |
|  | LIMIT | FREQUENCY | OF TOTAL |
|  | 1000 | 3000 | . 40 |
|  | 2000 | 21100 | 2.79 |
|  | 3000 | 12000 | 1.59 |
|  | 4000 | 9000 | 1.19 |
|  | 5000 | 15000 | 1.98 |
|  | 6000 | 11500 | 1.52 |
|  | 7000 | 10000 | 1.32 |
|  | 8000 | 10000 | 1.32 |
|  | 9000 | 10000 | 1.32 |
|  | 10000 | 7500 | . 99 |
|  | 11000 | 14600 | 1.93 |
|  | 12000 | 12000 | 1.59 |
|  | 13000 | 15000 | 1.98 |
|  | 14000 | 11500 | 1.52 |
|  | 15000 | 15500 | 2.05 |
|  | 16000 | 8600 | 1.14 |
|  | 17000 | 14600 | 1.93 |
|  | 18000 | 15000 | 1.98 |
|  | 19000 | 10500 | 1.39 |
|  | 20000 | 16000 | 2.11 |
|  | 21000 | 13500 | 1.78 |
|  | 22000 | 20000 | 2.64 |
|  | 23000 | 18500 | 2.44 |
|  | 24000 | 12500 | 1.65 |
|  | 25000 | 12100 | 1.60 |
|  | 26000 | 16000 | 2.11 |
|  | 27000 | 14500 | 1.92 |
|  | 28000 | 12500 | 1.65 |
|  | 29000 | 12000 | 1.59 |
|  | 30000 | 14000 | 1.85 |
|  | 31000 | 16000 | 2.11 |
|  | 32000 | 11500 | 1.52 |
|  | 33000 | 13000 | 1.72 |
|  | 34000 | 13000 | 1.72 |
|  | 35000 | 14000 | 1.85 |
|  | 36000 | 12000 | 1.59 |
|  | 37000 | 17500 | $2 \cdot 31$ |
|  | 38000 | 13000 | 1.72 |
|  | 39000 | 15000 | 1.98 |
|  | 40000 | 10000 | 1.32 |
|  | OVERFLOW | 234000 | 30.91 |

STANDARD DEVIATION
16662.293
16662.293

NON-WEIGHTED
761810.490

WEIGHTED

| CUMULATIVE PERCENTAGE | CUMULATIVE REMAINDER | MULTIPLE OF MEAN | DEVIATION <br> FROM MEAN |
| :---: | :---: | :---: | :---: |
| . 4 | 99.6 | . 033 | -1.746 |
| 3.2 | 96.8 | . 066 | -1.686 |
| 4.8 | 95.2 | . 100 | -1.626 |
| 6.0 | 94.0 | .133 | -1.566 |
| 7.9 | 92.1 | . 166 | -1.506 |
| 9.5 | $90 \cdot 5$ | .199 | -1.446 |
| 10.8 | 89.2 | . 233 | -1.386 |
| 12.1 | 87.9 | . 266 | -1.326 |
| 13.4 | 86.6 | . 299 | -1.266 |
| 14.4 | 85.6 | . 332 | -1.206 |
| 16.3 | 83.7 | . 365 | -1.146 |
| 17.9 | 82.1 | . 399 | -1.086 |
| 19.9 | 80.1 | . 432 | -1.026 |
| 21.4 | 78.6 | . 465 | -. 966 |
| 23.5 | 76.5 | . 498 | -. 906 |
| 24.6 | 75.4 | . 532 | -. 846 |
| 26.5 | 73.5 | . 565 | -. 786 |
| 28.5 | 71.5 | . 598 | -. 726 |
| 29.9 | $70 \cdot 1$ | . 631 | -. 666 |
| 32.0 | 68.0 | . 664 | -. 606 |
| 33.8 | 66.2 | . 698 | -. 546 |
| 36.4 | $63 \cdot 6$ | . 731 | -. 486 |
| 38.9 | 61.1 | . 764 | -. 426 |
| 40.5 | 59.5 | . 797 | -. 366 |
| 42.1 | 57.9 | . 831 | -. 306 |
| 44.3 | 55.7 | . 864 | -. 246 |
| 46.2 | 53.8 | . 897 | -. 186 |
| 47.8 | 52.2 | . 930 | -. 126 |
| 49.4 | 50.6 | . 963 | -. 066 |
| 51.3 | 48.7 | . 997 | -. 006 |
| 53.4 | 46.6 | 1.030 | . 054 |
| 54.9 | 45.1 | 1.063 | . 114 |
| 56.6 | 43.4 | 1.096 | . 174 |
| 58.3 | 41.7 | 1.130 | . 234 |
| 60.2 | 39.8 | 1.163 | . 294 |
| 61.8 | 38.2 | 1.196 | . 354 |
| 64.1 | 35.9 | 1.229 | . 414 |
| 65.8 | 34.2 | 1.263 | . 474 |
| 67.8 | 32.2 | 1.296 | . 534 |
| 69.1 | 30.9 | 1.329 | - 594 |
| 100.0 | - 0 |  |  |

TABLE NUMBER 82

| ENTRIES | IN TABLE 1488 | MEAN ARGUMENT -3948.972 |  |
| :---: | :---: | :---: | :---: |
|  | 757000 | -3850.649 |  |
|  | UPPER | OBSERVED | PERCENT |
|  | LIMIT | FREQUENCY | OF TOTAL |
|  | -20000 | 17500 | 2.31 |
|  | -19000 | 4000 | . 53 |
|  | -18000 | 4000 | - 53 |
|  | -17000 | 9000 | 1.19 |
|  | -16000 | 8600 | 1.14 |
|  | -15000 | 9100 | 1.20 |
|  | -14000 | 15100 | 1.99 |
|  | -13000 | 13000 | 1.72 |
|  | -12000 | 11000 | 1.45 |
|  | -11000 | 23000 | 3.04 |
|  | -10000 | 22100 | 2.92 |
|  | -9000 | 20000 | 2.64 |
|  | -8000 | 23000 | 3.04 |
|  | -7000 | 26000 | 3.43 |
|  | -6000 | 17500 | 2.31 |
|  | -5000 | 36000 | 4.76 |
|  | -4000 | 46500 | 6.14 |
|  | -3000 | 62500 | 8.26 |
|  | -2000 | 65500 | 8. 65 |
|  | -1000 | 61600 | 8.14 |
|  | 0 | 61000 | 8.06 |
|  | 1000 | 47000 | 6.21 |
|  | 2000 | 38500 | 5.09 |
|  | 3000 | 35000 | 4.62 |
|  | 4000 | 33500 | 4.43 |
|  | 5000 | 20500 | 2.71 |
|  | 6000 | 8500 | 1.12 |
|  | 7000 | 6500 | . 86 |
|  | 8000 | 4000 | - 53 |
|  | 9000 | 4000 | - 53 |
|  | 10000 | 1000 | -13 |
|  | 11000 | 500 | . 07 |
|  | 12000 | 500 | . 07 |
|  | 13000 | 500 | . 07 |
|  | 14000 | 0 | . 00 |
|  | 15000 | 500 | . 07 |
|  | 16000 | 0 | . 00 |
|  | 17000 | 0 | . 00 |
|  | 18000 | 500 | . 07 |
| REMAININ | G FREQUEN | ARE ALL |  |

REMAINING FREQUENCIES ARE ALL ZERO

| STANDARD DEVIATION 6415.291 |  | NON-WEIGHTED |  |
| :---: | :---: | :---: | :---: |
| 1661 |  | WEIGHTED |  |
| CUMULATIVE | CUMULATIVE | MULT IPLE | DEVIATION |
| PERCENTAGE | REMA INDER | OF MEAN | FROM MEAN |
| 2.3 | 97.7 | 5.065 | -2.502 |
| 2.8 | 97.2 | 4.811 | -2.346 |
| 3.4 | 96.6 | 4.558 | -2.190 |
| 4.6 | 95.4 | 4.305 | -2.034 |
| 5.7 | 94.3 | 4.052 | -1.878 |
| 6.9 | 93.1 | 3.798 | -1.723 |
| 8.9 | 91.1 | 3.545 | -1.567 |
| 10.6 | 89.4 | 3.292 | -1.411 |
| 12.1 | 87.9 | 3.039 | -1.255 |
| 15.1 | 84.9 | 2.786 | -1.099 |
| 18.0 | 82.0 | 2.532 | -. 943 |
| 20.7 | 79.3 | 2.279 | -. 787 |
| 23.7 | 76.3 | 2.026 | -. 631 |
| 27.1 | 72.9 | 1.773 | -. 476 |
| 29.4 | $70 \cdot 6$ | 1.519 | -. 320 |
| 34.2 | 65.8 | 1.266 | -. 164 |
| 40.3 | 59.7 | 1.013 | -. 008 |
| 48.6 | 51.4 | . 760 | . 148 |
| 57.3 | 42.7 | . 506 | . 304 |
| 65.4 | 34.6 | . 253 | . 460 |
| 73.4 | 26.6 | -. 000 | . 616 |
| 79.7 | 20.3 | -. 253 | . 771 |
| 84.7 | 15.3 | -. 506 | . 927 |
| 89.4 | 10.6 | -. 760 | 1.083 |
| 93.8 | 6.2 | -1.013 | 1.239 |
| 96.5 | 3.5 | -1.266 | 1.395 |
| 97.6 | 2.4 | -1.519 | 1.551 |
| 98.5 | 1.5 | -1.773 | 1.707 |
| 99.0 | 1.0 | -2.026 | 1.863 |
| 99.5 | - 5 | -2.279 | 2.018 |
| 99.7 | - 3 | -2.532 | 2.174 |
| 99.7 | - 3 | -2.786 | 2.330 |
| 99.8 | - 2 | -3.039 | 2.486 |
| 99.9 | -1 | -3.292 | 2.642 |
| 99.9 | -1 | -3.545 | 2.798 |
| 99.9 | -1 | -3.798 | 2.954 |
| 99.9 | -1 | -4.052 | 3.110 |
| 99.9 | . 1 | -4.305 | 3.265 |
| 100.0 | - 0 | -4.558 | 3.421 |

TABLE NUMBER 83
ENTRIES IN TABLE
1488
756500

UPPER
LIMIT
50
100
150
200
250
300
350
400
450
500
550
600
650
700
750
800
850
900
950
1000
1050
1100
1150
1200
1250
1300
1350
1400
1450
1500
1550
1600
1650
1700
1750
1800
1850
1900
1950
2000

| MEAN | $\begin{array}{r} \text { ARGUMENT } \\ 591.110 \end{array}$ |
| :---: | :---: |
|  | 580.221 |
| OBSERVED | PERCENT |
| FREQUENCY | OF TOTAL |
| 57600 | 7.61 |
| 64000 | 8.46 |
| 51000 | 6.74 |
| 37500 | 4.96 |
| 37000 | 4.89 |
| 36600 | 4.84 |
| 35100 | 4.64 |
| 33500 | 4.43 |
| 32000 | 4.23 |
| 27000 | 3.57 |
| 23600 | 3.12 |
| 31000 | 4.10 |
| 18000 | 2.38 |
| 23500 | 3.11 |
| 14500 | 1.92 |
| 23000 | 3.04 |
| 14000 | 1.85 |
| 24000 | 3.17 |
| 18000 | $2 \cdot 38$ |
| 20000 | 2.64 |
| 24000 | 3.17 |
| 13000 | 1.72 |
| 9000 | 1.19 |
| 9000 | 1.19 |
| 6500 | . 86 |
| 5100 | . 67 |
| 2500 | . 33 |
| 5000 | . 66 |
| 8500 | 1.12 |
| 6500 | . 86 |
| 5500 | - 73 |
| 3000 | . 40 |
| 1500 | - 20 |
| 2500 | - 33 |
| 2500 | - 33 |
| 2500 | - 33 |
| 1000 | - 13 |
| 1500 | - 20 |
| 3500 | . 46 |
| 3500 | . 46 |
| 19500 | 2.58 |

STANDARD DEVIATION
563.510
18083. 221

NON-WEIGHTED WEI GHTED

| CUMULAT IVE PERCENTAGE | CUMULATIVE REMAINDER | MULTIPLE OF MEAN | DEVIATION FROM MEAN |
| :---: | :---: | :---: | :---: |
| 7.6 | 92.4 | . 085 | -. 960 |
| 16.1 | 83.9 | . 169 | -. 872 |
| 22.8 | 77.2 | . 254 | -. 783 |
| 27.8 | 72.2 | . 338 | -. 694 |
| 32.7 | 67.3 | . 423 | -. 605 |
| 37.5 | 62.5 | . 508 | -. 517 |
| 42.1 | 57.9 | . 592 | -. 428 |
| 46.6 | 53.4 | . 677 | -. 339 |
| 50.8 | 49.2 | . 761 | -. 250 |
| 54.4 | 45.6 | . 846 | -. 162 |
| 57.5 | 42.5 | . 930 | -. 073 |
| 61.6 | 38.4 | 1.015 | . 016 |
| 64.0 | 36.0 | 1.100 | . 105 |
| 67.1 | 32.9 | 1.184 | . 193 |
| 69.0 | 31.0 | 1.269 | . 282 |
| 72.0 | 28.0 | 1.353 | . 371 |
| 73.9 | 26.1 | 1.438 | . 459 |
| 77.1 | 22.9 | 1.523 | . 548 |
| 79.4 | 20.6 | 1.607 | . 637 |
| 2?.1 | 17.9 | 1.692 | - 726 |
| 85.2 | 14.8 | 1.776 | . 814 |
| 87.0 | 13.0 | 1.861 | . 903 |
| 88.2 | 11.8 | 1.945 | . 992 |
| 89.3 | 10.7 | 2.030 | 1.081 |
| 90.2 | 9.8 | 2.115 | 1.169 |
| 90.9 | 9.1 | 2.199 | 1.258 |
| 91.2 | 8.8 | 2.284 | 1.347 |
| 91.9 | 8.1 | 2.368 | 1.435 |
| 93.0 | 7.0 | 2.453 | 1.524 |
| 93.9 | 6.1 | 2.538 | 1.613 |
| 94.6 | 5.4 | 2.622 | 1.702 |
| 95.0 | 5.0 | 2.707 | 1.790 |
| 95.2 | 4.8 | 2.791 | 1.879 |
| 95.5 | $4 \cdot 5$ | 2.876 | 1.968 |
| 95.8 | 4.2 | 2.961 | 2.057 |
| 96.2 | 3.8 | 3.045 | 2.145 |
| 96.3 | 3.7 | 3.130 | 2.234 |
| 96.5 | 3.5 | 3.214 | 2.323 |
| 97.0 | 3.0 | 3.299 | 2.411 |
| 97.4 | 2.6 | 3.383 | 2.500 |
| 100.0 | - 0 |  |  |

This program is the EXEC II version of GPSS II for processing on the UNIVAC 1107 computer. EXEC II permits GPSS II HELP Blocks and subroutines to be programmed in FORTRAN thus facilitating semi-professional programming as well as better understanding of the program contained in this appendix. GPSS II Program blocks begin on page 109. FORTRAN statements of reader interest include the Data Input Statements, page 99; the ten HELP subroutines, page 100; and subroutine FLOW and UPDATE, page 105.

## **** GLOSSARY *****

ARRAYS AND VARIABLES ASSOCIATED WITH JOB. CENTERS


ARRAYS AND VARIABLES ASSOCIATED WITH JOBS

```
JOB(I,J) ROWS ARE CENTER NO.S (ROUTE THRU SHOP FOR JOB I)
```

NEXT(I) INDEX J FOR NEXT CENTER IN JOB(I,J) FOR JOB I

```
Z(I) Z VALUE FOR JOB I. UPDATED AT END OF EACH DAY
NDUE(I) QUE DATE FOR JOB I
    JOBS ARE CLASSIFIED BY SIZE(NO.OF UNITS) AND BY INITIAL Z
    THE SIZE TYPE (1 TO 5) IS IN PARAMETER 4. TYPE 3 JOBS ARE
    TRACED.
    THE INITIAL & TYPE (1 TO 5) IS IN PARAMETER 3.
NTYPE(1). THE NO. OF SIZE TYPES
MAX 5
NTYPE(2) THE NO. OF INITIAL Z TYPES
    SIZE(K) THE NO. OF UNITS IN A JOB OF SIZE TYPE K
ZO(K) THE INITIAL Z FOR A JOB OF INIT. Z TYPE K
ISEED SEED NO.FOR RN GENERATOR USED TO DETERMINE ROUTE.
JOBIN(K,J) THE NO. OF JOBS ENTERING SHOP BY TYPE
                                    J=1 FOR SIZE TYPE K
                                    J=2 FOR ZO TYPE K
ARRAYS AND VARIABLES ASSOCIATED WITH PERIODIC STATUS REPORT.
DAY NO. OF CLOCK PERIODS IN A DAY.
NOAY NO. OF DAYS BETWEEN REPORTS.
            THE FOLLOWING ARRAYS ARE USED TO STORE STATISTICS
            ACCUMULATED UP THRU THE PREVIOUS REPORTING DAY.
JF(U) TIME CENTER J WAS IN USE.
NT(K) NO. OF ENTRIES IN TABLE K K=1.83
TM(K) SUM OF ENTRIES IN TABLE K
TSQ(K) SUM OF SQUARES OF ENTRIES IN TABLE K
UIN(K,J) THE NO. OF JOBS ENTERING SHOP BY TYPE
DIMENSION JF(20),MEANS(20),MEANP(20),TYPE(2),NTYPE(2),Z0(5),
1 SIZE(5),Z(501),NDUE (500),UIN(5,2):JOBIN( 5,2),SIZEZO(5,2)
DIMENSION FMEAN(5),FVAR(5)
COMMON/EQ1/JF1(.41)
COMMON/EQ2/JF2( 41)
COMMON/EQ3/JF3( 41)
COMMON/EQ4/JF4(41)
COMMON/EQ5/JF5( 41)
COMMON/EQ6/JF6( 41)
COMMON/QUE1/UQ1( 45)
COMMON/QUE2/JQ2( 45)
COMMON/QUE3/JQ3( 45)
COMMON/QUE4/JQ4( 45)
COMMON/QUE5/JQ5( 45)
COMMON/QUEG/JQ6( 45)
COMMON/TAB1/JTLOCS(100)
COMMON/TAB2/JTMODE(100)
COMMON/TAB3/JLOWRS(100)
COMMON/TAB4/JTINCS(100)
COMMON/TAB5/JTLAST(100)
COMMON/TAB6/UTLNUM(100)
COMMON/TAB7/TLARG (100)
COMMON/TAB8/TSQR (100)
COMMON/TAB9/TWARG (100)
COMMON/TAB10/TWSQR(100)
COMMON/EKSES/JEKS(50)
COMMON/CENTER/I,JOB(500,10),NEXT(500),CMEAN(20),CVAR(20),KMAX
COMMON/REPORT/K,NN(6),T(6),D(6),NT(83),TM(83),TSQ(83)
EQUIVALENCE (SIZEZO(1,1),SIZE(1)),(SIZEZO(1,2),ZO(1))
DATA NTRAN/500/.Z(501)/150./.DAY/1000./
```

```
DATA TYPE(1)/6HSZ= /,TYPE(2)/6HZO= /
```

DATA SWITCH/O./
THE FOLLOWING DATA IS VARIABLE INPUT DATA
the data caros may be changed
DATA ISEED /1354171/
DATA NCT/10/.KMAX/10/,KMIN/ 1/.SAMPLE/ 5./
DATA NDAY/5/
DATA NTYPE(1)/2//NTYPE(2)/1/
DATA (SIZE(J).J=1,5)/500.1100.1100.10.10.1
DATA (ZO(J), J=1.3)/0..0..0.1

DATA $\operatorname{MEANP}(J), J=1,10) / 1 \cdot 1 \cdot 1 \cdot 1 \cdot 1 \cdot 1,1,1 \cdot 1,1 /$
DATA (CMEAN(J),J=1,10)/500,500,500,500,500,500,500,500,500,500/

DATA (CVAR(J):J=1,10)/833.833.833.833.833.833.833.833.833.833/

DATA (FMEAN(J),J=1,5)/500,100,100,0.0/

DATA (FVAR(J),J=1.5)/833.33.33.0.0/
C

500 FORMAT(I4,F6.2.I6.I5.2F9.2.1X2F9.2,I5,F7.2.I5,I7.F7.2.1X.10I3/
1 90X10I3)
501 FORMAT (4X.F6.2.I6.I5.2F9.2.1X.2F9.2/)
502 FORMAT (1X,A3,F7.2,15, 15,2F9.2,1X,2F9.2)
503 FORMAT(10X,I6,I5.2F9.2.1X.2F9.2/)
504 FORMAT(10H ALL TYPES.IG.I5.2F9.2.1X.2F9.2.4X2F9.2)
505 FORMAT(10X,I6.I5.2F9.2.1X.2F9.2.4X.2F.9.2/)
506 FORMAT (24H1****** REPORT FOR LAST II3.6H DAYS..9X4HDATE,I4//
$121 H$ CTR UTIL JOBS JOBS, $4 \times 4 \mathrm{HFLOW} 5 \times 4 \mathrm{HFLOW} 6 \times 4 H W A I T 5 X 4 H W A I T 3 X 2 H I N$
$15 \times 1 \mathrm{HZ}$
2 4X3HCUR2X3HDUE6X1HZ4X7HCENTERS/2X3HNO.8X2HIN3X3HOUT4X.4HMEAN4X
3 6HST. DEV5X4HMEAN4X6HST.DEV2X3HQUE2X4HNEXT3X3HJOB2X4HDATE1OX
4 4HLEFT )
507 FORMAT(2X3HJOB7X9HJOBS JOBS $4 \times 4 \mathrm{HFLOW} 5 \times 4 \mathrm{HFLOW6X3HD-A6X3HD-A10X4HEXIT}$
1 5X4HEXIT/2X4HTYPE7X2HIN3X3HOUT4X4HMEAN4X6HST.DEV5X4HMEAN4X
2 6HST.DEV8X4HMEAN4X6HST.DEV)
701 FORMAT (/11H *ENTER JOBI4.17H INTO SHOP CLOCKI7.2X3HD-CIG.2X2HZ= 1F6.2,2X4HSIZE,F6.0,1X3HDUEI7,2X5HROUTE,10I3/90X,10I3)
702 FORMAT(/ 11H *SEND JOBI4, 7H TO QUEI3.2X5HCLOCKI7.2X3HD-CI6.2X 12HZ=F6.2.3X6HT MEANF8.2.2X6HST.DEVF8.2.3X6HQ MEANF8.2.2X6HST.DEV 2F8.2)
703 FORMAT(/ 11H *EXIT JOBI4.17H FROM SHOP CLOCKI7.2X3HD-AI6.2X2HZ= 1F6.2.2X4HSIZE,F6.0.1×3HDUEI7)
706 FORMAT(/ 11H \#CHK $Z$ JOBI4, 7H IN QUEI3. $2 \times 5 \mathrm{HCLOCKI712} \mathrm{\times 3HD-CI6.2X}$ 12HZ=F6.2.2X9HZ IN CTR=F6.2.2X7HNEXT Z=F7.2)
708 FORMATG 11H *SETUP JOBI4. 7H IN CTRI3.2X5HCLOCKI7.2X3HD-CI6.2X
12HZ=F6.2.3X 6HS MEANI5.5X6HSPREADI5)
709 FORMATE 11H *PROC. JOBI4, 7H IN CTRI3,2X5HCLOCKI7,2X3HD-CI6.2X
12HZ=F6.2.3X 6HP MEANI5.5X6HSPREADI5)
1003 FORMAT(/30H RESET TABLES TO ZERO AT CLOCK , I7/)

THERE ARE TEN ENTRY POINTS AT STATEMENTS 1 THRU 10
GO TO (1,2,3,4,5,6,7,8,9,10):JX
$1 . I=J F i(41\rangle$
C STORE TRANS. NO IN SAVEX 41 JEKS(41)=I
C: THE NO. OF CENTERS FOR THIS JOB =KMAX-KCT+I' UNIFORMLY DISTRIBUTED
C. FROM KMIN TO KMAX.

CALL RANDM(NEWSED•RN)

```
        KCT=RN*SPREAD+1.
        NEXT(I)=KCT
C ASSIGN CENTERS UNIFORMLY FROM 1 TO NCT.
        DO 12 J=KCT,KMAX
        CALL RANDM(NEWSEDIRN)
    12 JOB(I,J)=XNCT*RN+1.
    13. Z(I)=Z0(JZ)
C CALCULATE DUE DATE FOR JOB I.
        XC=FLOAT (KMAX-KCT+1)
        XN=FLOAT(UTLNUM(JMOD+60))
        IF(XN.LT.SAMPLE) GO TO 14
        SMEAN=TLARG (JMOD+60)/XN
        SVAR=TSQR(JMOD+60)/XN-SMEAN*SMEAN
        FMEAN(UMOD)=2.*SMEAN/(XNCT+1.)
        FVAR(UNVD)=2.*SVAR/(XNCT+1.)
    14: NDUE(I)= Z(I)*SQRT(ABS(FVAR(JMOD)*XC))+FMEAN(JMOD)*XC+FLOAT(JMEAN)
        UOBIN(JMOD,1)=JOBIN(JMOD,1)+1
        JOBIN(JZ,2)=JOBIN(JZ,2)+1
        IF(JMOD.NE,3) RETURN
C
C WRITE TRACE LINE FOR JOBS OF SIZE TYPE 3.
    NDC=NDUE (I)-JMEAN
    WRITE(6.701)I,JMEAN,NDC,Z(I), SIZE(JMOD),NDUE(I),(JOB(I,J),
    1 J=KCT,KMAX)
    RETURN
    HELP2 OBTAINS THE NEXT CENTER FOR THIS JOB.
    IF JOB IS FINISHED NBA IS BLOCK 28, OTHERWISE NBA= BLOCK 15
    2 KCT=NEXT(UZ)
    STORE NBA IN SAVEX 42. STORE NEXT CENTER NO. IN SAVEX 43
    JEKS(42)=28
    IF(KCT.GT.KMAX) RETURN
    JEKS(43)=JOB(JZ,KCT)
    JEKS(42)=15
    IF(JMOD.NE.3) RETURN
C WRITE TRACE LINE FOR JOBS OF SIZE TYPE 3.
    J=JEKS(43)
    NDC=NDUE(JZ)-JMEAN
    DEV=SQRT(ABS(CVAR(J)))
    IF(JTLNUM(J+40).EQ.O) GO TO 21
    SMEAN=TLARG(J+40)/FLOAT(JTLNUM(J+40))
    SVAR=SQRT(ABS (TSQR(J+40)/FLOAT (JTLNUM(J+40))-SMEAN*SMEAN))
    WRITE(6,702) JZ,J,UMEAN,NDC,Z(JZ),CMEAN(J),DEV,SMEAN,SVAR
    RETURN
21.WRITE(6,702) JZ,J.JMEAN,NDC,Z(JZ),CMEAN(J),DEV
    RETURN
C
C
C
    3.JEKS(44)=NDUE(JZ)-JMEAN
    IF(JMOD.NE.3) RETURN
C
C WRITE TRACE LINE FOR JOBS OF SIZE TYPE 3.
    NDC=NOUE(JZ)-JMEAN
```

```
    WRITE(6,703) JZ,JMEAN,NDC,Z(JZ), SIZE(JMOD),NDUE(JZ)
    RETURN
C
    4 DO 41 J=1.NCT
    XN=FLOAT(JTLNUM(J))
C USE INITIAL ESTIMATES UNTIL THE NO.OF JOBS COMPLETED IN THIS
C CENTER = SAMPLE.
    IF(XN.LT.SAMPLE)GO TO 41
C FLOW TIME FOR CENTER JIS TABULATED IN TABLE J
C JTLNUM, TLARG, TSQR = NO.OF ENTRIES, SUM OF ENTRIES, SUM OF
C : SQUARES OF ENTRIES IN TABLE.
    CMEAN(J)=TLARG(J)/XN
    CVAR(J)=TSQR(J)/XN-CMEAN(J)*CMEAN(J)
    41 CONTINUE
C RE-CALCULATE Z VALUE.
        DO 43 I=1.NTRAN
        IF(NEXT(I).GT.KMAX) GO TO 43
        CALL FLOW(I,SMEAN,SVAR)
    42 Z(I)=(FLOAT(NDUE(I)-JZ)-SMEAN)/SVAR
    43 CONTINUE
        NDATE=JZ/IFIX(DAY)
        C1=JZ
        JEKS(46)=5
        IDAY=IDAY+1
C TEST FOR END OF REPORTING PERIOD.
        IF(IDAY.GE.NDAY) JEKS(46)=80
        RETURN
C
    5 CALL RSTART(ISEED)
        XNCT=FLOAT(NCT)
        JEKS(50)=NCT
        XC1=DAY*FLOAT (NDAY)
        SPREAD=KMAX-KMIN+1
        DO 50 I=1,NTRAN
        50 NEXT(I)=KMAX+1
        DO 51 I=1.83
        NT(I)=0
        TM(I)=0.
52 TSQ(I)=0.
        DO 52 I=1,NCT
52 JF(I)=0
        DO 53 I=1,5
        DO 53 J=1,2
        JOBIN(I,U)=0
53 JIN(I,J)=0
        IDAY=0
        RETURN
C
    HELPG ASSIGNS AN INTEGER PRIORITY ACCORDING TO THE JOBS Z VALUE
```

C

```
    6 JEKS(45)=15000-IFIX(100.*Z(JZ))
    IF(JMOD.NE,3) RETURN
C WRITE TRACE LINE FOR JOBS OF SIZE TYPE 3.
    NDC=NDUE(JZ)-JMEAN
    KCT=NEXT (JZ)
    J=JOB(JZ,KCT)
    JC=JF1(J)
    IF CENTER IS IDLE, SET DUMMY JOB NO.=501, Z(501)=150.
    IF(JC.LE.0) JC=501
    ZN=.01*FLOAT(15000-JEKS(J))
    WRITE(6,706) JZ:J,JMEAN,NDC,Z(JZ):Z(JC):ZN
    RETURN
```

C
HELP7 OUTPUTS THE STATUS REPORT EVERY NDAY DAYS.
HELP7 IS ENTERED ONCE FOR EACH CENTER VIA TRANSACTION CONTROL LOOP
WRITE HEADING FOR STATUS REPORT.
IF (IDAY.NE.O) WRITE (6,506) NDAY,NDATE
IDAY=0
DO $71 \mathrm{~K}=1.2$
I=TABLE NO. FLOW TIME FOR CENTER J IS IN TABLE J
QUE TIME FOR CENTER $ل$ IS IN TABLE $40+J$
$I=40 *(K-1)+J Z$
CALL UPDATE(K,I)
C SUBROUTINE UPDATE CALCULATES THE MEAN AND STD.DEV OF THE VARIABLE
C IN TABLE NO. I FOR THE REPORTING PERIOD AND FOR THE TOTAL TIME TO
C
71 CONTINUE
$N Q=J M E A N+J M O D$
UTIL=FLOAT (JF2 (JZ+20)-JF (JZ))/XC1
$J F(J Z)=J F 2(J Z+20)$
$I=J F 1(J Z)$
QZ=.01*FLOAT(15000-JEKS (JZ))
IF(I.EQ.O)GO TO 73
KCT=NEXT(I)
WRITE (6,500) JZ.UTIL,NN(2),NN(1),(T(K),D(K),K=1,2),NQ,QZ,I,NDUE(I),
$12(I),(J O B(I, J), J=K C T, K M A X)$
GO TO 74
73 WRITE(6.500)JZ,UTIL,NN(2),NN(1),(T(K),D(K),K=1,2),NQ
74 UTIL=FLOAT (JF2 (JZ+20))/C1
WRITE (6.501) UTIL,NN(5),NN(4), (T(K),D(K),K=4,5)
$\operatorname{JEKS}(46)=80$
C TEST FOR LAST CENTER. IF FINISHED NBA=BLOCK 5, OTHERWISE NBA=80
IF (JZ.LT.NCT) RETURN
C
$\operatorname{JEKS}(46)=5$
WRITE 6,507 )
DO $76 \mathrm{~K} 1=1,2$
K3 $=$ NTYPE (K1)
DO $76 \mathrm{~K} 2=1 . \mathrm{KJ}$
DO $75 \mathrm{~K}=1.2$
$I=60+10 *(K 1-1)+5 *(K-1)+K 2$
$I=T A B L E$ NO. FLOW TIME FOR SIZE TYPE J IS IN TABLE 60+J
D-A TIME FOR SIZE TYPE J IS IN TABLE 65+J
FLOW TIME FOR 20 TYPE لIS IN TABLE 70+ل

| C 75 | D－A TIME FOR ZO TYPE $J$ IS IN TABLE 75＋J |
| :---: | :---: |
|  | CALL UPDATE（K，I） |
|  | CONTINUE |
|  | IN＝JOBIN（K2，K1）－JIN（K2，K1） |
|  | $\operatorname{JIN}(\mathrm{K} 2, \mathrm{~K} 1)=\mathrm{JOBIN}(\mathrm{K} 2 \cdot \mathrm{~K} 1)$ |
|  |  |
| 76 | WRITE（6，503）JOBIN（K2，K1），NN（4），（T（K），D（K），K＝4，5） |
|  | IN＝JF6（41）－NJF6 |
|  | NJF6＝JF6（41） |
|  | DO $77 \mathrm{~K}=1.3$ |
|  | $\mathrm{I}=80+\mathrm{K}$ |
| C | I＝TABLE NO．FLOW TIME FOR ALL JOBS IS IN TABLE 81 |
| C | D－A TIME FOR ALL JOBS IS IN TABLE 82 |
| C | INTER EXIT TIME FOR ALL JOBS IN TABLE 83 |
|  | CALL UPDATE（K，I） |
| 77 | CONTINUE |
|  | WRITE（6，504）IN，NN（1），（T（K），D（K），K＝1，3） |
|  | WRITE（6，505）NJF6 ，NN（4），（T（K），D（K），$K=4,6)$ |
|  | RETURN |
| C |  |
| C |  |
| C | HELP8 PUTS MEAN SETUP TIME FOR CENTER（JMEAN）IN SAVEX 42，AND |
| $C$ PUTS SPREAD IN SAVEX 43 |  |
| C |  |
| 8 | JEKS（42）＝MEANS（JMEAN） |
|  | JEKS（43）＝JEKS（42）／10 |
|  | IF（JMOD．NE．3）RETURN |
| C |  |
| C | WRITE TRACE LINE FOR JOBS OF SIZE TYPE 3. |
|  | NDC＝NDUE（JZ）－JF3（JMEAN） |
|  | WRITE（6，708）JZ，JMEAN，JF3（JMEAN），NDC，Z（JZ），JEKS（42），JEKS（ 43 ） |
|  | RETURN |
| C |  |
| $c$ |  |
| C | HELPg PuTS MEAN PROCESS TIME IN SAVEX 42 AND SPREAD IN SAVEX 43 |
| C |  |
| 9 | JEKS（42）$=$ MEANP（ JMEAN）＊IFIX（SIZE（JMOD）） |
|  | JEKS（43）$=\operatorname{JEKS}(42) / 10$ |
|  | NEXT $(J Z)=$ NEXT $(J Z)+1$ |
|  | IF（JMOD．NE．3）RETURN |
| C |  |
| C | WRITE TRACE LINE FOR JOBS OF SIZE TYPE 3. |
|  | NDC＝NDUE（JZ）－JEKS（49） |
|  | WRITE（6，709）JZ．JMEAN，JEKS（49），NDC，Z（JZ），JEKS（42），JEKS（43） |
|  | RETURN |
| $C$$C$ |  |
|  |  |
| C | HELP10 RESETS TABLES TO ZERO |
| C HELP10 RESETS TABLES TO ZERO |  |
| 10 | D0 $1001 \mathrm{I}=1.83$ |
|  | $N T(I)=0$ |
|  | TM（I）$=0$ 。 |
|  | TSQ（I）$=0$ 。 |
|  | $\operatorname{JTLNUM}(\mathrm{I})=0$ |
|  | TWARG（I）$=0$ 。 |
|  | TWSQR（I）$=0$ 。 |
|  | $\operatorname{TLARG}(I)=0$. |
| 100 | $T S Q R(I)=0$ ． |

```
WRITE(6.1003) JZ
RETURN
END
SUBROUTINE FLOW CALCULATES MEAN AND ST.DEV. OF FLOW TIME THRU
CENTERS REMAINING FOR JOB I.
SUBROUTINE FLOW(I,SMEAN,SVAR)
COMMON/CENTER/I,JOB(500,10),NEXT(500),CMEAN(20),CVAR(20),KMAX
N1=NEXT(I)
SMEAN=0.
SVAR=0.
DO 101 K=N1.KMAX
JCTR=JOB(I,K)
SMEAN=SMEAN+CMEAN(JCTR)
SVAR=SVAR+CVAR(JCTR)
SVAR=SQRT(ABS(SVAR))
RETURN
END
NN(K)=JTLNUM(I)-NT(I)
T(K)=TLARG(I)-TM(I)
D(K)=TSQR(I)-TSQ(I)
IF(NN(K).EQ.O) GO TO 201
XNT=NN(K)
NT(I)=JTLNUM(I)
TM(I)=TLARG(I)
TSQ(I)=TSQR(I)
T(K)=T(K)/XNT
D(K)=SQRT(ABS(D(K)/XNT-T(K)*T(K)).)
NN(K+3)=JTLNUM(I)
T(K+3)=0.
D(k+3)=0.
IF(JTLNUM(I).EQ.O) RETURN
XNT=JTLNUM(I)
T(K+3)=TLARG (I)/XNT
D(K+3)=SQRT (ABS(TSQR(I)/XNT-T(K+3)*T(K+3)))
RETURN
END
SUBROUTINE UPDATE CALCULATES THE MEAN AND STD.DEV OF THE VARIABLE IN TABLE NO. I FOR THE REPORTING PERIOD AND FOR THE TOTAL TIME TO DATE
SUBROUTINE UPDATE(K•I)
I=TABLE NO.
\(K=I N D E X\) FOR TEMP STORAGE ARRAYS NN(K),T(K).D(K) WHERE NN,T.D ARE THE NO. OF ENTRIES:MEAN, AND DEVIATION
        K=1 FOR FLOW TIME OVER REPORT PERIOD
        K=2 FOR D-A TIME OVER REPORT PERIOD
        K=3 FOR EXIT TIME OVER REPORT PERIOD
    K=4 FOR FLOW TIME OVER TOTAL PEERIOD
    K=5 FOR D-A TIME OVER TOTAL PERIOD
    K=6 FOR EXIT TIME OVER TOTAL PERIOD
NT(I)=TM(I),TSQ(I) ARE THE NO. OF ENTRIES, SUM OF ENTRIES, AND THE
        SUM OF SQUARES OF ENTRIES IN TABLE NO I AT THE END OF THE
        PREVIOUS REPORTING PERIOD.
COMMON/REPORT/K,NN(6),T(6),D(6),NT(83),TM(83),TSQ(83)
COMMON/TABG/JTLNUM(100)
COMMON/TAB7/TLARG(100)
COMMON/TAB8/TSQR(100)
```



```
    COMMON/STOR4/JS4(1)
    COMMON/STOR5/JS5(1)
    COMMON/STOR6/JS6(1)
    COMMON/STOR7/JS7(1)
C
    QUEUE GOMMON/QUE1/JQ1( 45
    COMMON/QUE2/JQ2( 45)
    COMMON/QUE3/JQ3( 45)
    COMMON/QUE4/JQ4( 45)
    COMMON/QUE5/JQ5( 45)
    COMMON/QUE6/JQ6( 45)
C LOGIC SWITCH TABLE
    COMMON/LOGIX/JL1(25)
C SAVEX TABLE
    COMMON/EKSES/JEKS(50)
C FUNCTION TAB
    COMMON/FN1/JYLOCS(5)
    COMMON/FN2/JXLOCS(5)
    COMMON/FN3/JSLOCS(5)
    COMMON/FN4/JFNPAN(5)
    TABLE AND QTABLE TABLES NEXT 10 CARDS DIMENSION IDENTICALLY
    COMMON/TAB1/JTLOCS(100)
    COMMON/TAB2/JTMODE(100)
    COMMON/TAB3/ULOWRS(100)
    COMMON/TAB4/JTINCS(100)
    COMMON/TAB5/JTLAST(100)
    COMMON/TAB6/JTLNUM(100)
    COMMON/TAB7/TLARG (100)
    COMMON/TAB8/TSQR (100)
    COMMON/TAB9/TWARG (100)
    COMMON/TAB10/TWSQR(100)
C VARIABLE STATEMENT TABLE
    COMMON/VARS/JVLOCS(10)
C COMMON CORE AREA
                                1 CARD
    COMMON/WORDS/JWORDS(4500)
C TRANSACTION TABLES NEXT 9 CARDS DIMENSION IDENTICALLY
    COMMON/TRAN1/JNDT(500)
    COMMON/TRAN2/JCHAIN(500)
    COMMON/TRAN3/JMOVE(500)
    COMMON/TRAN4/JNNWD(500)
    COMMON/TRAN5/JC1(500)
    COMMON/TRAN6/JC2(500)
    COMMON/TRAN7/JC3(500)
    COMMON/TRAN8/JC4(500)
    COMMON/TRAN9/JC5(500)
C DO NOT REDIMENSION ANY OF THE FOLLOWING CARDS
    COMMON/TRAN10/JC6(1)
    COMMON/TRAN12/JC7(1)
    COMMON/TRAN14/JC8(1)
    COMMON/TRAN16/JC9(1)
    COMMON/TRAN18/JC10(1)
    COMMON/TRAN20/JC11(1)
    COMMON/TRAN22/JC12(1)
    COMMON/TRAN24/JC13(1)
    COMMON/TRAN26/JC14(1)
    COMMON/TRAN28/JC15(1)
    COMMON/TRAN30/JC16(1)
    COMMON K(100)
```

```
    COMMON LPRI(8)
    COMMON LPRIOR(8),ICHAR(70),KTYPE(41),KGATE(12),KCONTR(7),KSV(17),
    1 KCOMP(6),KSELEC(7),LX(6)
    OIMENSION FWORDS(1)
    EQUIVALENCE(JWORDS(1),FWORDS(1))
    EQUIVALENCE(K(1),KASYM1),(K(2),KASYM2),(K(3),KNODES),(K(4),KEQS),
    1 (K(5),KSTORS),(K(6),KQUES),(K(7),KVARS),(K(8),KLOGIX),
    2 (K(9),KEKSES),(K(10),KFNS),(K(11),KTABS),(K(12),KWORDS),
    3 (K(13),KTRANS),(K(14),KRAND),(K(15),KASMBL),(K(16),KIT),
    4 (K(17),KOT)
    EQUIVALENCE(K(78),KPARAM),(K(71),INDFLO),(K(72),INDEND),
    1 (K(55),IFATAL)
            KPARAM = 8
    KIT = 5
    KOT = 6
    KRAND = 1220703125
    KNODES = MDIFF(JN2(1),JN1(1))
    KEQS = MDIFF(JF2(1),JF1(1))
    KSTORS = MDIFF(JS2(1),JS1(1))
    KQUES = MDIFF(JQ2(1),JQ1(1))
        KVARS = MDIFF(JWORDS(1),UVLOCS(1))
        KLOGIX = MDIFF(JEKS(1),لL1(1))
        KEKSES = MDIFF(JYLOCS(1),JEKS(1))
    KFNS = MDIFF(JXLOCS(1),JYLOCS(1))
    KTABS = MDIFF(JTMODE(1),JTLOCS(1))
        KWORDS = MDIFF(JNDT(1),JWORDS(1))
    KTRANS = MDIFF(JCHAIN(1).JNDT(1))
        CALL BLOCKD
        CALL INPROC($20,$30)
    IF (INOFLO .NE. O) CALL FLOW
    IF (IFATAL •NE. O) GO TO 10
    IF (INDEND .NE. 0) GO TO 10
        CALL EXECUT
        CALL PUTOUT
        GO TO 10
30 CALL ASSEMB
    GO TO 10
    END
```







VITA
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Candidate for the Degree of
Doctor of Philosophy

Thesis: THE EFFECT OF ORDER SIZE ON THE OPERATION OF A HYPOTHETICAL JOB SHOP MANUFACTURING SYSTEM

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