# COMPUTER AIDED OPTIMIZATION OF NONLINEAR SERVOMECHANISMS EMPLOYING A drected search of multiparameter component librarles and statistical. TOL.Erancing 

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## This is to certify that the

## thesis entitled

Computer Aided Optimization of Nonlinear Servomechanisms Employing A Directed Search of Multiparameter Component Libraries And Statistical Tolerancing.
presented by

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## ABSTRACT

COMPUTER AIDED OPTIMIZATION OF NONLINEAR SERVOMECHANISMS EMPLOYING A DIRECTED SEARCH OF MULTIPARAMETER COMPONENT LIBRARIES and Statistical tolerancing<br>by Bruce Allen Chubb

Techniques are developed to automatically select from computerized libraries the components that satisfy a given system specification at minimum dollar cost. This is accomplished by defining an object function to be the system cost which in turn is a function of the probability that the design will be successful in meeting the performance specification. Starting with an initial set of component part numbers, the total system cost is minimized by iterating the component part numbers using a directed search technique. The manufacturing tolerances associated with the component parameters are considered in calculating the probability of success.

Comparisons are made between the Monte Carlo and the directed search procedure which illustrate that the directed search technique has considerable advantage. Several examples demonstrate that such a computer program can result in considerable cost savings.

The techniques are developed around an instrument servomechanism as a specific example. Four component libraries are established to list the part characteristics for the followup, amplifier,
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assigned performa:
and time response
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step inputs, and
motor-generator, and geartrain. Combinations of up to eight preassigned performance specifications in the areas of damping, accuracy, and time response are considered. The nonlinear effects of backlash, coulomb friction, and amplifier saturation are considered as well as the effect of finite geartrain stiffness in evaluating the system performance. Equations are derived for calculating, 1) the allowable backlash without a limit cycle, 2) the nonlinear overshoot for large step inputs, and 3) the effective bandwidth for sinusoidal inputs.

# COMPUTER AIDED OPTIMIZATION OF <br> NONLINEAR SERVOMECHANISMS EMPLOYING A <br> DIRECTED SEARCH OF MULTIPARAMETER COMPONENT LIBRARIES <br> AND STATISTICAL TOLERANCING 

By

Bruce Allen Chubb

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Finally the
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## 1. INTRODUCTION

### 1.1 STATEMENT OF THE PROBLEM

The system engineer operating within the framework of a typical manufacturing organization operates from the following basic information and constraints :
a) A set of customer specifications to be met,
b) A basic system configuration to be used in realizing these specifications,
c) A set of standard components that fit into this configuration.

The basic problem is to determine the collection of components that satisfies the given specification at minimum total dollar cost.

Automated techniques for selecting the optimum set of components for a system are necessitated by today's competitive market and the multitude of candidate components available. As an example, consider the problem of selecting an optimum set of components for a fixed amplifier configuration to meet a given set of customer specifications. If one extrapolates data from a 1964 survey [1], today there should be approximately 60,000 semiconductor devices manufactured and designated by part number. If one adds to this, the number of standard resistors, capacitors, transformers, etc., it becomes obvious
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that manual techniques cannot come close to yielding an optimum component selection.

The same situation exists in every area of system engineering where the configuration is "fixed" and a multitude of candidate components are available. The characteristics of these components can, of course, be stored in computer libraries by part numbers and an analysis program can be written to systematically analyze the system for any candidate set of components by merely inserting the appropriate part numbers. Such computer programs are structured so as to retrieve the data for each particular component, proceed with the various performance calculations and display the results to the designer for each set of part numbers manually selected.

The goal of this study is to go one step further and develop techniques, procedures, and programs for the effective use of computers in automating the solution to the above class of design problem.

### 1.2 EXAMPLE SYSTEM

The techniques presented are developed around an instrument servomechanism as a specific example. The design problem is essentially the same as that discussed in references [2,3]; however, the techniques developed are believed to be much improved.

The example instrument servomechanism consists of a follow-up device, electronic amplifier, drive motor with feedback generator, and geartrain. A pictorial diagram showing a fixed system configuration using these components is shown in Figure 1.1.


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Figure 1.1. Schematic diagram of motor-generator instrument servomechanism.

It is assumed that the design of this configuration must meet up to eight preassigned specifications in the areas of damping, accuracy, and time response. Four component libraries are established to list the part characteristics as follows:
a) Follow-up - 25 part numbers
b) Amplifier - 50 part numbers
c) Motor-generator - 25 part numbers
d) Geartrain - 25 part numbers

Even though the size of each demonstration library was purposely kept small, the number of theoretical possible candidate systems is large, namely: $25 \times 50 \times 25 \times 25=781,250$.

The optimum collection of components is defined as "the one that satisfies the given specification in a manner resulting in minimum total cost".

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Three ancillary problems considered in the thesis have their origin in the fact that design equations did not heretofore exist for calculating the allowable backlash, large step overshoot, and bandwidth for a nonlinear instrument system. Solutions to these three problems represent significant advancements in the field of servomechanism design, and are presented as Appendices.

### 1.3 SURVEY OF PRESENT TECHNIQUES

Literally thousands of articles have been published which list and describe work that has been done in optimization. A few of the more comprehensive publications are listed as references $[4,5,6]$. Most of this work is concerned with finding that set of $n$ parameter values, $X_{1}, X_{2}, \cdots, X_{n}$, which maximizes (or minimizes) a given scalar function $F\left(X_{1}, X_{2}, \cdots, X_{n}\right)$, subject to constraints on these parameters which limit their range to realizable values.

Three of the most popular techniques are centered about the following three basic approaches:
a) Random Experimentation Although crude forms of this method are as old as design technology itself, the best early formal documentation as given in 1958 [7] uses repeated solution of the system design equations with random selections of the input parameters generated through Monte Carlo methods. In its simple form, a large number of computer solutions are required to achieve good results. When for reasons of computer costs, only a few runs can be justified, a partitioned
who
or stratified form [8] usually provides better efficiency. In general, improved results are obtained most often if some form of strategy or learning can be employed to adjust the frequency distributions representing the parameters to be selected.
b) Steepest Ascent

This method was introduced by R. R. Brown in 1957 [9] and further improved in 1959 [10]. Today there are many computer programs available for general use which employ this technique. The Steepest Ascent methods calculate the partial derivatives $\partial F / \partial X_{1}, \partial F / \partial X_{2}, \cdots, \partial F / \partial X_{n}$ usually numerically, and then proceed along the gradient until a maximum is obtained. Since the result represents only a local or relative maximum, various starting points are used in an attempt to find the global maximum. GREAT [11] is one example of a highly effective program that is based on this method.
c) Direct Search

The Direct Search technique is attributed to Hook and Jeeves who presented the unconstrained case in 1961 [12]. This was modified in 1965 by Weisman and Wood [13] to include constraints. In direct search, the minimum is found by the sequential examination of a finite set of trial values. The result of each trial is compared with the best previous trial and the new value accepted if an improvement is observed. This series of exploratory moves, in which each
variable is individually adjusted, is used to determine the "best" direction for a successful move. This move, in which all parameters are changed, is called the "pattern move'. Each pattern move is followed by a sequence of exploratory moves to revise the pattern. The sequence is repeated until the scalar function can be increased no further. A good application of this type of algorithm is LOOK[14].

### 1.4 SCOPE OF INVESTIGATION

A logic flow diagram representing an effective design procedure is shown in Figure 1.2. As indicated, there are two design iteration loops. One loop concerns changes in the basic system configuration (e.g. component interconnection) and the other changes in component selection. Only the problems associated with automating the component selection are considered here.

If each component could be represented by a single parameter, the solution would be quite straightforward. One could arrange the components in the library in ascending order of its single parameter and use a modified form of either the steepest ascent or direct search method to find the optimum. However, the general solution is far more difficult, since libraries consist of multiterminal components, and several parameters are required to describe each component. These parameters are associated only with the component part number, and there is no natural ordering between components.



Figure 1.2. A simplified design flow diagram.

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The problem is analogous to A. M. Gleason's Search in the $N$-Cube [15]. Gleason states "It is entirely clear that there is no certain method of finding the maximum, short of computing the function at each point of the set in question". This exhaustive search for our particular design problem, however, is out of the question. For example, if we consider 30 seconds to be required for each analysis using a high speed machine, it would take 6510 hours of computer time to solve the problem in question. There must be a better way!

The problem can be divided into three aspects. The first aspect is concerned with developing an effective analysis program, including formulation of the necessary nonlinear state equations, codification of the system specifications, developing the required design equations and a method of handling the component parameter tolerances. This effort is presented in Section 2.

The second task is to incorporate the analysis program in the design loop by adding an optimization procedure. To this end it is necessary to formulate the object function to be minimized, set up component libraries, and to formulate the optimization strategy. This effort is presented in Section 3.

Section 4 of this thesis presents the application of the developed program to typical hardware design problems. Section 5 then presents the conclusions of the study and provides suggested guidelines for future work.

## 2. DEVELOPMENT OF ANALYSIS PROGRAM

### 2.1 BASIC APPROACH

The analysis section is the starting point of any computer-aided or automated design program. Optimization, in the design context, is derived from an efficient use of iterative analysis techniques. Devoid of a good analysis capability, the designer has nothing. Its presence provides a powerful tool in itself. In this case, however, it is simply a means to an end - Automated Design.
"But what are the requirements for an effective analysis program?" First, and primary, is the fact that it must accurately represent the hardware. This requires a significantly detailed model, including often overlooked nonlinearities, and a realistic consideration of component tolerance effects. This means that the programmer is faced with the solution of nonlinear differential equations, and that system parameters, instead of being constants, must be treated as random variables. Second the outputs of the analysis program must have a one-to-one correspondence with the list of system specifications. That is, if the customer specifies overshoot, response time, accuracy, etc., then the program must have the capability of calculating the performance characteristics in this form. Third and last, since the analysis is to be repeated many times in an iterative fashion, the solution time should be a minimum.

### 2.2 FORMULATION OF SYSTEM STATE EQUATIONS

A most effective method of obtaining the response of a system is by using the state variable model [16]. Much work has been done in the effective application of this approach to the analysis of physical systems $[17,18]$. Many aspects of the particular problem considered here are presented in reference [2]. However, in the interest of continuity a limited development is repeated here.

The example system under consideration consists of a followup, amplifier, servomotor with an integral mounted feedback generator, geartrain, and load. The load is made up of inertia and coulomb friction. Experience had demonstrated [2] that geartrain resilience, along with the nonlinearities of gear backlash, amplifier saturation, and coulomb friction, must be considered.

The state model diagram of the system is shown in Figure 2.1 and the system and component parameters are defined in Tables 2.1 and 2.2 , respectively. Four state variables are required to define the system. These are the outputs of the 4 integrators of Figure 2.1 and correspond to motor velocity, motor position, load velocity, and load position. It should be noted that the motor velocity and position have been reflected to equivalent values at the load (i.e. the hardware values are simply those given by Figure 2.1 times the gear ratio, $N$ ). The amplifier saturation is represented by an equivalent torque saturation (i.e. the torque level is set at a value equal to that of the amplifier voltage level times the product of the motor torque gain and gear ratio).

Figure 2.1. Nonlinear state model diagram for an instrument servomechanism.

Table 2

Symbol
$B_{M}$
$J_{L}$
$J_{M}$
$K_{i}$
$K_{T}$
$T_{L}$
$\mathrm{T}_{\text {SAT }}$

Table 2.

Table 2.1. System parameter definitions for state model.

| Symbol | Defining Equation | Name | Units |
| :---: | :---: | :---: | :---: |
| $B_{M}$ | $N^{2} \mathrm{~B}_{\mathrm{m}}$ | Reflected motor damping | 02-in/rad/sec |
| $J_{L}$ | $\mathbf{J}_{\ell}+\mathbf{J}_{\mathbf{g}}$ | Load inertia | oz-in/rad/sec ${ }^{2}$ |
| $\mathbf{J}_{\mathbf{M}}$ | $\mathrm{N}^{2} \mathrm{~J}_{\mathrm{m}}$ | Reflected motor inertia | $\mathrm{oz}-\mathrm{in} / \mathrm{rad} / \mathrm{sec}^{2}$ |
| $\mathrm{K}_{\mathrm{G}}$ | $K_{g} K_{a g} K_{m} N^{2}$ | Generator damping coefficient | oz-in/rad/sec |
| $K_{T}$ | $K_{f} K_{a f} K_{m} N$ | System torque constant | oz-in/rad |
| $T_{L}$ | $T_{\ell}+T_{g}$ | Load friction | oz-in |
| $\mathrm{T}_{\text {SAT }}$ | $K_{m} E_{\text {sat }}$ | System torque saturation | 02-in |

Table 2.2. Component parameter definitions for state model.

| Symbol | Definition | Units |
| :---: | :---: | :---: |
| B | Geartrain backlash | radians |
| $\mathrm{B}_{\text {m }}$ | Motor viscous damping | oz-in/rad/sec |
| $E_{\text {sat }}$ | Amplifier saturation level | volts |
| $\mathrm{J}_{8}$ | Geartrain inertia | oz-in/rad/sec ${ }^{2}$ |
| $\mathbf{J}_{\ell}$ | Load inertia | oz-in/rad/sec ${ }^{2}$ |
| $\mathbf{J}_{\text {m }}$ | Motor-generator inertia | 02-in/rad/sec ${ }^{2}$ |
| $K_{\text {af }}$ | Auplifier gain to followup | volts/volt |
| $K_{\text {ag }}$ | Amplifier gain to generator | volts/volt |
| $K_{f}$ | Followup gain | volts/rad |
| $K_{g}$ | Generator gain | volts/rad/sec |
| ${ }_{\mathbf{K}}$ | Motor torque gain | oz-in/volt |
| $K_{s}$ | Geartrain spring constant | oz-in/rad |
| $N$ | Gear ratio |  |
| $\mathrm{T}_{\mathrm{g}}$ | Geartrain friction | oz-in |
| $\mathrm{T}_{\boldsymbol{\ell}}$ | Load coulomb friction | 02-in |



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The nonlinear functions representing backlash ( $N_{1}$ ), coulomb friction $\left(N_{2}\right)$, and saturation $\left(N_{3}\right)$ are defined in Figure 2.2.

| BLOCK diagram | EQUATIONS(Where: $\operatorname{sgn} \theta=\theta / 1 \theta \mid$ ) |
| :---: | :---: |
|  | $\begin{aligned} & T_{r} \cdot K_{L}\left[\left\|\theta_{u}-\theta_{L}\right\|-B\right] \operatorname{son}\left(\theta_{u}-\theta_{L}\right) \\ & \text { for }\left\|\theta_{m}-\theta_{L}\right\| \geq B \\ & T_{r}: 0 \\ & \text { for }\left\|\theta_{\mu}-\theta_{L}\right\| \leq B \end{aligned}$ |
| $\mathrm{N}_{2}$ | $\begin{aligned} & T_{t}=\left\lvert\, \begin{array}{l} T_{L} \mid \operatorname{son} \dot{\theta}_{L} \\ \text { for } \dot{\theta}_{L} \end{array}=0\right. \\ & T_{f}=-T_{r} \\ & \quad \text { for } \dot{\theta}_{L}=0 \end{aligned}$ |
|  | $\begin{aligned} & T_{m}=K_{T}\left(\theta_{1}-\theta_{L}\right)-K_{0} \dot{\theta}_{M} \\ & \quad \text { for }\left\|K_{T}\left(\theta_{1}-\theta_{L}\right)-K_{0} \dot{\theta}_{u n}\right\| \leq T_{B A T} \\ & T_{m}=T_{\text {mar }} \sin \left[K_{T}\left(\theta_{1}-\theta_{L}\right)-K_{0} \dot{\theta}_{M}\right] \\ & \quad \text { for }\left\|K_{T}\left(\theta_{1}-\theta_{L}\right)-K_{0} \dot{\theta}_{M}\right\| \geq T_{\text {Bar }} \end{aligned}$ |

Figure 2.2. Nonlinear function definitions.

The nonlinear state model for the system can be obtained directly from Figure 2.1 as:
$\frac{d}{d t}\left[\begin{array}{c}\dot{\theta}_{M} \\ \theta_{M} \\ \dot{\theta}_{L} \\ \theta_{L}\end{array}\right]=\left[\begin{array}{cccc}-\frac{B_{M}}{J_{M}} & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0\end{array}\right]\left[\begin{array}{c}\dot{\theta}_{M} \\ \theta_{M} \\ \dot{\theta}_{L} \\ \theta_{L}\end{array}\right]+\left[\begin{array}{l}\frac{1}{J_{M}}\left(T_{m}-T_{r}\right) \\ 0 \\ \frac{1}{J_{L}}\left(T_{r}-T_{f}\right) \\ 0\end{array}\right]$

The cor
$N_{1}=1$ for
$N_{2}=1$ for
$\frac{d}{d t}\left[\begin{array}{l}\dot{\theta}_{M} \\ \theta_{M} \\ \dot{\theta}_{L} \\ \theta_{L}\end{array}\right]$

In the speci
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The corresponding linear approximation is obtained by setting $N_{1}=1$ for zero backlash; $N_{2}=0$ for zero coulomb friction; and $N_{3}=1$ for no amplifier saturation.

$$
\frac{d}{d t}\left[\begin{array}{c}
\dot{\theta}_{M}  \tag{2.2}\\
\theta_{M} \\
\dot{\theta}_{L} \\
\theta_{L}
\end{array}\right]=\left[\begin{array}{cccc}
-\frac{\left(B_{M}+K_{G}\right)}{J_{M}} & -\frac{k_{S}}{J_{M}} & 0 & \frac{\left(k_{S}-K_{T}\right)}{J_{M}} \\
1 & 0 & 0 & 0 \\
0 & \frac{k_{s}}{J_{L}} & 0 & -\frac{k_{S}}{J_{L}} \\
0 & 0 & 1 & 0
\end{array}\right]\left[\begin{array}{l}
\dot{\theta}_{M} \\
\theta_{M} \\
\theta_{M} \\
\dot{\theta}_{L} \\
\theta_{L}
\end{array}\right]+\left[\begin{array}{l}
\frac{k_{T}}{J_{M}} \\
0 \\
0
\end{array}\right]
$$

In the special case when the geartrain stiffness is considered infinite (i.e. $K_{s}=\infty$ ) the linear state model becomes

$$
\frac{d}{d t}\left[\begin{array}{c}
\dot{\theta}_{L} \\
\theta_{L}
\end{array}\right]=\left[\begin{array}{cc}
-f_{T} /{ }_{T} & -\mathrm{K}_{T} /{ }^{J} T \\
1 & 0
\end{array}\right]\left[\begin{array}{l}
\dot{\theta}_{L} \\
\theta_{L}
\end{array}\right]+\left[\begin{array}{c}
\mathrm{K}_{T} / J_{T} \\
0
\end{array}\right]{ }_{\theta_{i}}
$$

where $f_{T}$ is the total effective viscous damping from the feedback generator and motor. This is

$$
\begin{equation*}
\mathrm{f}_{\mathrm{T}}=\mathrm{B}_{\mathrm{M}}+\mathrm{K}_{\mathrm{G}} \tag{2.4}
\end{equation*}
$$

and in a like manner $J_{T}$ is the total system inertia given by

$$
\begin{equation*}
\mathrm{J}_{\mathrm{T}}=\mathrm{J}_{\mathrm{M}}+\mathrm{J}_{\mathrm{L}} \tag{2.5}
\end{equation*}
$$

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Equation (2.3) may be written also in the convenient form

$$
\frac{d}{d t}\left[\begin{array}{c}
\dot{\theta}_{L}  \tag{2.6}\\
\theta_{L}
\end{array}\right]=\left[\begin{array}{cc}
-2 \zeta \omega_{N} & -\omega_{N}{ }^{2} \\
1 & 0
\end{array}\right]\left[\begin{array}{l}
\dot{\theta}_{L} \\
\theta_{L}
\end{array}\right]+\left[\begin{array}{c}
\omega_{N}{ }^{2} \\
0
\end{array}\right]{ }^{\theta_{i}}
$$

where $\zeta$ and $\omega_{\mathrm{N}}$ are the damping ratio and natural frequency as normally defined for a second order system. For the particular case under consideration

$$
\begin{align*}
\zeta & =\frac{f_{T}}{2 \sqrt{K_{T} J_{T}}}  \tag{2.7}\\
\omega_{N} & =\sqrt{\frac{K_{T}}{J_{T}}} \tag{2.8}
\end{align*}
$$

### 2.3 SYSTEM SPECIFICATIONS AND DESIGN EQUATIONS

The first step in realizing a design is to establish a thorough understanding of the set of performance specifications that the system must satisfy. The second step required is to develop a set of equations that enable one to evaluate a potential design in relation to the specifications. This section is devoted to the accomplishment of both the above tasks.

For a computer program to be effective in design, it must cover a somewhat general set of specifications. Then, for any given application, the user may choose the particular desired set and instruct the computer to ignore the others. A set of eight specifications is
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selected for the example program developed as part of this study. They are representative of those listed in numerous military and commercial specifications for such systems as manufactured by the Instrument Division of Lear Siegler, Incorporated, Grand Rapids, Michigan.

The eight specifications are now discussed one at a time, with the corresponding design equations used to evaluate a proposed design.

1) Static Accuracy

Static accuracy is unquestionably the most often specified requirement for any instrument servo. It is simply a measure of the magnitude of the error that can exist between the command input and the indicated output of the system under static conditions. Contributions to this error include followup tracking error, amplifier and generator null offsets, motor starting voltage, and gearing and load friction. By taking each of these error sources and dividing by the corresponding dc gain back to the error angle, and summing, the following equation is derived for the static accuracy ( $\varepsilon_{A}$ ).

$$
\begin{equation*}
\varepsilon_{A}={ }^{\theta_{f}}+\frac{E_{a n}}{K_{f} K_{a f}}+\frac{K_{a g} E_{g n}}{K_{f} K_{a f}}+\frac{E_{s}}{K_{f} K_{a f}}+\frac{T_{g}+T_{\ell}}{K_{f} K_{a f} K_{m} N} \tag{2.9}
\end{equation*}
$$

where

$$
\begin{aligned}
\theta_{f}= & \text { followup tracking accuracy (rad) } \\
\mathrm{E}_{\mathrm{an}}= & \text { amplifier output null voltage (volts) } \\
\mathrm{E}_{\mathrm{gn}}= & \text { generator output null voltage (volts) } \\
\mathrm{E}_{\mathrm{s}}= & \text { motor-generator no-load starting voltage } \\
& \text { (volts) }
\end{aligned}
$$

and all other notation is defined in Table 2.2.
2) Resolution

Resolution is a measure of the total dead-zone in an instrument servomechanism. It therefore represents the maximum amount that the input can be displaced without noting any motion at the output. This deadzone results from the fact that a certain amount of error must be built-up to overcome the motor starting voltage and coulomb frictions. Thus, the total deadzone or resolution $\left(\varepsilon_{R}\right)$ is given by

$$
\begin{equation*}
\varepsilon_{R}=2\left[\frac{E_{s}}{K_{f} K_{a f}}+\frac{T_{g}+T_{\ell}}{K_{f} K_{a f} K_{m} N}\right] \tag{2.10}
\end{equation*}
$$

3) Velocity Lag

Velocity lag is a measure of the servo's accuracy under constant velocity conditions. It is defined as the steadystate positional difference between the command input and the indicated output with the input rotating at a constant velocity. Since the resulting lag error is a function of the input velocity, the latter also must be specified.

The velocity lag ( $\varepsilon_{L}$ ) may be calculated using the equation (see reference [2])

$$
\begin{equation*}
\varepsilon_{L}=\left[\frac{N^{2}\left(B_{m}+K_{g} K_{a g} K_{m}\right)}{K_{f} K_{a f} K_{m}^{N}}\right] \dot{\theta}_{i n}+\varepsilon_{A} \tag{2.11}
\end{equation*}
$$

where $\dot{\theta}_{\text {in }}$ is the input velocity at which the lag error is to be measured or calculated and $\varepsilon_{A}$ in the static accuracy as defined by (2.9).
4) Followup Rate

Followup rate is a measure of the maximum velocity that the servo is capable of producing. If there were no friction loading, it would be simply the motor no-load speed divided by the gear ratio. However to account for the load, one can calculate the followup rate ( $\dot{\theta}_{L}$ ) using the equation: (see reference [2])

$$
\begin{equation*}
\dot{\theta}_{L}=\frac{\dot{\theta}_{m}}{N}\left[1-\frac{T_{g}+T_{\ell}}{N T_{s}}\right] \tag{2.12}
\end{equation*}
$$

where the symbols are as defined in Table 2.2 except for the additional ones which are

$$
\begin{aligned}
& \dot{\theta}_{m}=\text { motor no-load speed }(\mathrm{rad} / \mathrm{sec}) \\
& \mathrm{T}_{\mathrm{s}}=\text { motor stall torque }(\mathrm{oz}-\mathrm{in})
\end{aligned}
$$

5) Damping Ratio

Damping ratio is the most often used measure of system
stability. This is unfortunate since its definition applies only for a linear 2nd order system. However if one makes this linear approximation, then the damping ratio equation may be obtained directly in terms of the component parameters by substituting the definitions of Table 2.1 into Equation (2.7). Thus:

$$
\begin{equation*}
\zeta=\frac{N^{2}\left(B_{m}+K_{g} K_{a g} K_{m}\right)}{2 \sqrt{K_{f} K_{a f} K_{m} N\left(N^{2} J_{m}+J_{g}+J_{\ell}\right)}} \tag{2.13}
\end{equation*}
$$

6) Null Oscillation

Null oscillations are small amplitude steady state oscillations (limit cycles) that exist about a null and are a result of backlash being present in the geartrain. A typical specification states that "no such oscillation shall exist." In Reference [2], it was established that the amount of backlash that a given design can tolerate without such a limit cycle is proportional to the amount of coulomb friction on the load side of the backlash. In this study, we shall derive the equation for the proportionality constant (derivation in Appendix A) thereby obtaining the equation for the allowable backlash as follows:
$B($ allowable $)=M 1 \frac{N}{0<\omega<\sqrt{\frac{L_{T}}{J_{L}}}}\left\{\frac{3.2}{\pi\left[N_{1}(\omega)+.4\right] \omega N_{2}(\omega)} \sqrt{\frac{\left(K_{T}-J_{M} \omega^{2}\right)^{2}+\left(f_{T} \omega\right)^{2}}{\left(N_{1}(\omega) K_{s}-J_{M} \omega^{2}\right)^{2}+\left(f_{T} \omega\right)^{2}}}\right\} T_{L}$

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where:



As long as the actual backlash $B$ is less than $B(a l l o w a b l e)$
a limit cycle will not exist.
7) Overshoot

Overshoot of the system's output to a step command input is the most often used measure of servo response. If the servo is linear and of second order, the overshoot is defined by the damping ratio ( $\zeta$ ) given by the equation (see Reference [2])

Overshoot $=\left[e^{-\pi \zeta / \sqrt{1-\zeta^{2}}}\right] \theta_{\text {step }} \quad$ for $\zeta<1$

$$
\begin{equation*}
=0 \text { for } \zeta \geq 1 \tag{2.17}
\end{equation*}
$$

However, because of system nonlinearities, mainly amplifier saturation and coulomb friction, the size of the actual overshoot is not proportional to the step size and is not given by a simple relationship such as (2.17). The actual overshoot could be obtained by a direct simulation of the nonlinear state model, Equation (2.1), however such a numerical solution is quite time consuming on a digital computer. For this reason, the nonlinear state equations are solved explicitly (see Appendix B for solution), thereby
enabling a much more direct calculation for the overshoot. This is accomplished by using piecewise linear solutions over the regions shown in the phase plane diagram of Figure 2.3. This illustrates a response trajectory of a typical system and the corresponding overshoot. As can be seen, there are three regions of operation. In region 1 , the servo has negative torque saturation, while in region 2 the servo is unsaturated, and finally in region 3 there is positive torque saturation. The solutions for the system state vector, as derived in Appendix B for each region, are summarized in Table 2.3. By solving the first equation at the saturation boundary, using the result as initial conditions for the appropriate second equation, and again finding the next boundary conditions, one can proceed from boundary point to boundary point along the trajectory until the overshoot is obtained. The actual logic used to obtain the boundary conditions is summarized in Appendix B.
8) Bandwidth

Bandwidth is a measure of the systems ability to follow sinusoidally oscillating inputs. It is normally defined as the frequency at which the output response is attenuated to 0.707 times the input ( -3 db ). For a linear second order system this bandwidth frequency $\left(\omega_{B}\right)$ is given by

$$
\begin{equation*}
\omega_{B}=\omega_{N} \sqrt{1-2 \zeta^{2}+\sqrt{2-4 \zeta^{2}+4 \zeta^{4}}} \tag{2.18}
\end{equation*}
$$

where $\omega_{\mathrm{N}}$ is as defined in (2.8)

Figure 2.3. Phase plane diagram showing piecewise linear regions.

Table 2.3. State equations for calculating nonlinear overshoot.

Region 1 (Negative torque saturation)


Region 2 (Unsaturated)



Region 3 (Positive torque saturation)

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However, because of system nonlinearities; namely, saturation and coulomb friction, the actual system bandwidth is a function of the amplitude of the input sinusoid. The necessary procedure for including this nonlinear effect is developed as part of this study. The development is included as Appendix $C$ and is based on the use of describing function approximations to obtain effective values for $\omega_{N}$ and 5.

The eight system specifications that have now been described are summarized in Table 2.4. This table lists the name, symbol, and number assigned to each specification, tells whether the specification is an upper or lower bound, and the units used. In addition to specifying any desired combination of the above described eight performance requirements, the user must also define the load that the servo is to drive. For the example program developed as part of this study, the load is represented by an inertia ( $J_{\ell}$ ) and a nonlinear coulomb friction ( $\mathrm{T}_{\ell}$ ).

Table 2.4. System specifications.

| Name | Symbol | Boundary | Units |
| :--- | :---: | :---: | :---: |
| Static accuracy | $\mathrm{S}_{1}$ | upper | degrees |
| Resolution | $\mathrm{S}_{2}$ | upper | degrees |
| Velocity lag | $\mathrm{S}_{3}$ | upper | degrees |
| Follow-up rate | $\mathrm{S}_{4}$ | lower | deg/sec |
| Damping ratio | $\mathrm{S}_{5}$ | lower | - |
| Allowable backlash | $\mathrm{S}_{6}$ | upper | minutes |
| Overshoot | $\mathrm{S}_{7}$ | upper | degrees |
| Bandwidth | $\mathrm{S}_{8}$ | lower | hertz |

The analysis problem can be now defined mathematically by letting $S, Y$, and $X$ be vectors, defined in general as:

$$
\begin{align*}
S & =\left[S_{1}, S_{2}, \cdots, S_{k}\right] \\
Y & =\left[Y_{1}, Y_{2}, \cdots, Y_{k}\right] \\
X & =\left[X_{1}, X_{2}, \cdots, X_{n}\right] \tag{2.19}
\end{align*}
$$

where

$$
\begin{aligned}
k= & \text { number of performance specifications } \\
n= & \text { number of component parameters } \\
S_{i}= & \text { numerical value for the } i^{\text {th }} \text { specification as } \\
& \text { defined in Table } 2.4(1 \leq i \leq k) \\
Y_{i}= & \text { system performance function corresponding to } \\
& i^{\text {th }} \text { specification }(1 \leq i \leq k) \\
X_{j}= & \text { numerical value for } j^{\text {th }} \text { component parameter } \\
& (1 \leq j \leq n)
\end{aligned}
$$

Thus one can write in general that

$$
\left[\begin{array}{l}
Y_{1}  \tag{2.20}\\
Y_{2} \\
\vdots \\
Y_{k}
\end{array}\right]=\left[\begin{array}{l}
F_{1}\left(X_{1}, X_{2}, X_{3}, \cdots, X_{n}\right) \\
F_{2}\left(X_{1}, X_{2}, X_{3}, \cdots, X_{n}\right) \\
\vdots \\
F_{k}\left(X_{1}, X_{2}, X_{3}, \cdots, X_{n}\right)
\end{array}\right]
$$

It is only necessary, at this time, that the $X$ vector contain the elements as required to calculate the system performance function vector $Y$. However, it is convenient to include the component costs as part of the $X$ vector [even though they will not show up explicitly in (2.20)] since they are required to calculate the
optimization function that is introduced later. Following this practice for our particular example, $k=8$ and $n=23$, where the $X$ vector is defined in Table 2.5. Likewise, the $F$ functions are defined in Table 2.6.

Table 2.5. Component parameter definitions for library.

| COMP | VAR | PARMMETER NAME | SYMBOL | UNITS |
| :---: | :---: | :---: | :---: | :---: |
| F $\mathbf{O}$ $\mathbf{L}$ $\mathbf{L}$ $\mathbf{L}$ $\mathbf{W}$ $\mathbf{U}$ $\mathbf{P}$ | $\begin{aligned} & x_{1} \\ & x_{2} \\ & x_{3} \end{aligned}$ | Cost <br> Gain <br> Accuracy | $\begin{aligned} & \mathbf{c}_{\mathbf{f}} \\ & \mathbf{k}_{\mathbf{f}} \\ & { }_{\theta_{\mathbf{f}}} \end{aligned}$ | dollars <br> volts/rad <br> minutes |
| A M $\mathbf{P}$ $\mathbf{L}$ $\mathbf{I}$ $\mathbf{F}$ I $\mathbf{E}$ R | $\begin{aligned} & x_{4} \\ & x_{5} \\ & x_{6} \\ & x_{7} \\ & x_{8} \\ & \hline \end{aligned}$ | Cost <br> Gain to Followup <br> Gain to Generator <br> Output Saturation Level Output Null Voltage | $\begin{aligned} & \mathrm{C}_{\mathrm{a}} \\ & \mathrm{~K}_{\mathrm{af}} \\ & \mathrm{~K}_{\mathrm{ag}} \\ & \mathrm{E}_{\mathrm{sat}} \\ & \mathrm{E}_{\mathrm{an}} \\ & \hline \end{aligned}$ | dollars <br> volts/volt <br> volts/volt <br> volts <br> volts |
|  | $\begin{aligned} & \hline x_{9} \\ & x_{10} \\ & x_{11} \\ & x_{12} \\ & x_{13} \\ & x_{14} \\ & x_{15} \end{aligned}$ | Cost <br> Stall Torque <br> No-load Speed <br> Inertia <br> Starting Voltage <br> Generator Gain <br> Generator Null | $\begin{aligned} & \mathrm{C}_{\mathrm{m}} \\ & \mathrm{~T}_{\mathrm{s}} \\ & \dot{\theta}_{\mathrm{m}} \\ & \mathrm{~J}_{\mathbf{m}} \\ & \mathrm{E}_{\mathbf{s}} \\ & \mathrm{K}_{\mathrm{g}} \\ & \mathrm{E}_{\mathrm{gn}} \end{aligned}$ | dollars <br> oz-in <br> rpm <br> $\mathrm{gm}-\mathrm{cm}^{2}$ <br> volts <br> volts/1000 rpm <br> millivolts |
|  | $\begin{aligned} & x_{16} \\ & x_{17} \\ & x_{18} \\ & x_{19} \\ & x_{20} \\ & x_{21} \end{aligned}$ | Cost <br> Inertia <br> Stiffness <br> Friction <br> Backlash <br> Gear Ratio | $\begin{aligned} & \mathrm{C}_{\mathrm{g}} \\ & \mathrm{~J}_{\mathrm{g}} \\ & \mathrm{~K}_{\mathrm{S}} \\ & \mathrm{~T}_{\mathrm{g}} \\ & \mathrm{~N}^{2} \end{aligned}$ | dollars <br> $\mathrm{gm}-\mathrm{cm}^{2}$ <br> oz-in/rad <br> oz-in <br> minutes |
| $\begin{aligned} & 0 \\ & \text { A } \\ & \text { D } \end{aligned}$ | $\mathrm{X}_{22}$ $\mathrm{X}_{23}$ | Inertia Priction | $\begin{aligned} & \mathrm{J}_{\boldsymbol{\ell}} \\ & \mathrm{T}_{\boldsymbol{\ell}} \end{aligned}$ | $\begin{aligned} & \mathrm{gm}-\mathrm{cm}^{2} \\ & 02-\mathrm{in} \end{aligned}$ |

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Table 2.6. Location of $F$ Functions.

| Function | Location |
| :---: | :--- |
| $\mathbf{F}_{\mathbf{1}}$ | Equation (2.9) |
| $\mathbf{F}_{\mathbf{2}}$ | Equation (2.10) |
| $\mathbf{F}_{\mathbf{3}}$ | Equation (2.11) |
| $\mathbf{F}_{\mathbf{4}}$ | Equation (2.12) |
| $\mathbf{F}_{\mathbf{5}}$ | Equation (2.13) |
| $\mathbf{F}_{\mathbf{6}}$ | Actual backlash (X ${ }_{20}$ ) |
| $\mathbf{F}_{\mathbf{7}}$ | Table 2.3 with logic from <br> Figure B.3 |
| $\mathbf{F}_{\mathbf{8}}$ | Equation (2.18) with $\omega_{\mathrm{N}}$ and <br> $\quad$replaced by effective values <br> as defined in Appendix C |

Thus (2.20) can be used to calculate the system performance vector ( Y ) given any component vector ( X ). By programming this equation as presented, one obtains the desired analysis program except for one deficiency. That is, due to manufacturing tolerances, the $X$ vector varies from unit to unit, and we are interested not in a particular value of $Y$ but what spread or limits to expect. For this reason, the next section is devoted to selecting a suitable method for determining this tolerance spread.

### 2.4 VARIABILITY ANALYSIS TECHNIQUES

Variability Analysis refers to the methods used to determine the ability of a system to continue to give specified performance while its component parts change value within specified limits.

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One method of insuring that a given system design meets all of the specified performance critera is to use some form of worst-case analysis. An example of this type of procedure is MANDEX which is a worst-case circuit analysis computer program [19]. Using this program the first derivative of all the output variables with respect to each of the input parameters is used to set each of the parameters to their "worst-case" tolerance extreme, so that a "worst-case" condition exists at each of the circuit outputs.

For multivariable systems, the application of the worst-case method becomes very time consuming, even when using a high speed computer, due to the multitude of possibilities that must be considered. Even if the worst-case stackup can be found, the resulting design is unrealistic since it assumes that everything is at worst-case on the same system at the same time. Using this criteria consistently throughout the whole design invariably results in component tolerance requirements that are so tight the cost is prohibitive. The resulting system is greatly overdesigned.

The application of statistical tolerance theory to iterative design problems overcomes this difficulty and provides a most realistic picture of the control system behavior to be expected in production. Statistical tolerance theory was first introduced by Shewhart in his book "Economic Control of Quality of Manufactured Products" [20]. Following this, S. S. Wilks of Princeton University in the early 1940's published two papers [21], [22] that developed the statistical foundation for tolerance theory. However these articles and those that followed [23], [24], [25] up until as late
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as 1963 concerned themselves almost universally with the design problem of assigning tolerances to the physical dimensioning of mechanical parts. From a systems point of view, this case applies only to the situation where the system function is a linear combination of the component parameters. In general, and for this example, system performance is a complex nonlinear function of the component parameters and the simple root-sum-square technique is not adequate. The Monte Carlo and Moment methods developed in the last few years may be applied to handle this problem. Both techniques are presented and the merits of each are compared by D. G. Mark [26] and by Mark and L. H. Stember [27].

The Monte Carlo technique assumes that each component parameter under investigation can be represented by a frequency distribution. A system is simulated mathematically by choosing each parameter value randomly from its frequency distribution. After parameter values for each component in the system are selected, a solution is obtained. Then each parameter value is again chosen as before and another solution is obtained. This sequence is repeated many times, resulting in a tabulation of data representing the distributions of the desired output variables. From this, the resulting mean and 3 sigma values can be calculated.

The Moment technique makes use of an expansion of the function about the mean parameters using a Taylor series. The higher order terms of the series are usually neglected. This requires taking the partial derivative of each performance variable with respect to each component parameter. Assuming that the component performance
parameters are independent and noting that the $\partial Y_{i} / \partial X_{j}=0$ if $X_{j}$ is a component cost, the mean value of $Y_{i}$ is given by the equation

$$
\begin{equation*}
\left.Y_{i}(\text { mean })=F_{i}\left[X_{1} \text { (mean }\right), X_{2}(\text { mean }), \cdots, X_{n} \text { (mean) }\right] \tag{2.21}
\end{equation*}
$$

and the standard deviation of $Y_{i}$ is approximated by the equation:

$$
\begin{equation*}
\sigma_{Y_{i}}=\sqrt{\left[\left(\sigma_{X_{1}}\right) \frac{\partial Y_{i}}{\partial X_{1}}\right]^{2}+\left[\left(\sigma_{X_{2}}\right) \frac{\partial Y_{i}}{\partial X_{2}}\right]^{2}+\cdots+\left[\left(\sigma_{X_{n}}\right) \frac{\partial Y_{i}}{\partial X_{n}}\right]^{2}} \tag{2.22}
\end{equation*}
$$

where $i=1,2, \cdots, k$ and the partial derivatives are evaluated while all other parameters are held at their mean value.

Since the higher order derivatives are neglected, the Moment method prediction is considered less accurate than the Monte Carlo method, but still adequate for most purposes. The Moment method has the advantage that it provides information that is extremely useful to the designer in pinpointing sensitive areas and reducing this sensitivity to parameter variability. Because of this latter advantage and the fact that satisfactory results can be obtained with a lesser number of computer runs, the Moment method is used here.

As can be seen from (2.22), the use of the Moment method requires that we calculate the partial derivatives of each system performance function with respect to each component parameter. The matrix of these partials is the Jacobian

$$
J=\frac{\partial\left(Y_{1}, Y_{2}, \cdots, Y_{k}\right)}{\partial\left(X_{1}, X_{2}, \cdots, X_{n}\right)}=\left[\begin{array}{ccc}
\frac{\partial Y_{1}}{\partial X_{1}} & \frac{\partial Y_{1}}{\partial X_{2}} \cdots & \frac{\partial Y_{1}}{\partial X_{n}}  \tag{2.23}\\
\frac{\partial Y_{2}}{\partial X_{1}} & \frac{\partial Y_{2}}{\partial X_{2}} \cdots & \frac{\partial Y_{2}}{\partial X_{n}} \\
\vdots & \vdots & \vdots \\
\frac{\partial Y_{k}}{\partial X_{1}} & \frac{\partial Y_{k}}{\partial X_{2}} \cdots & \frac{\partial Y_{k}}{\partial X_{n}}
\end{array}\right]
$$

Approximation of these partials is easily obtained numerically by programming (2.20) and using a subroutine to make the following steps:

1) Set all the $X_{i}$ 's equal to their mean value $\left(\bar{X}_{i}\right)$, and the calculated $Y$ vector is taken to be the mean value $\bar{Y}$.
2) $X_{1}$ is replaced by $\left(\bar{X}_{1}+\Delta X_{1}\right)$ and the corresponding value of $Y$ is calculated with all other $X$ 's at their mean value. From this, we obtain the first column of the Jacobian matrix using
$\frac{\partial Y_{i}}{\partial X_{j}} \simeq \frac{Y_{i}-\bar{Y}_{i}}{\Delta X_{j}} \quad$ for $\quad i=1,2, \ldots, k$ and $j=1$
3) Step 2 is repeated for each $X_{j}$ for $J=1,2, \cdots$, n thereby obtaining the complete Jacobian matrix.
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## 3. DEVELOPMENT OF COMPUTER OPTIMIZATION DESIGN PROCEDURE

### 3.1 BASIC APPROACH

Use of the computer-aided design procedure developed in the previous section, although many times more effective than any manual method, nevertheless represents only a passive use of the digital computer. That is, the engineer makes all the design decisions and the computer only serves as a fast calculator. The next logical step toward optimized design is to use the computer to determine how the components should be varied to converge on the desired minimum cost system.

Figure 3.1 illustrates in general, how a computer could be used in a dynamic sense. The prerequisite to design is to input the data for all components. This is accomplished by loading in the component data cards pre-punched in a prescribed format. This need be done only the first time and thereafter only if that data is to be changed, e.g., updated. These data are then stored by part number in an easily retrievable form on magnetic disk and are referred to as the "component libraries." In order to provide the mainline design program with a guide as to part number selection, some ordered array of these is desired. This is accomplished by using a "search matrix library," the precise working of which is explained later. Thus, immediately


Figure 3.1. Computer aided design program flow chart.
after generation of the component libraries, the computer calculates the component search matrices and stores these in a second block of data -- the search matrix library. Now the program is ready to be used. The designer inputs the system specifications, fixed production labor cost, and any initial set of components of his choice. The latter item could be made a random selection if desired. In either event, the computer retrieves the component data from libraries and proceeds to calculate the system performance. The component parameters are then perturbated one at a time and the partials of each system performance function with respect to each component parameter are determined. Once this is completed the partials are stored in the form of a Jacobian matrix. The calculated performance limits are then compared to the specification limits. The fraction of the units produced that statistically fall outside of the specification limits is then calculated as the "rejection ratio." From this rejection ratio, the fixed labor cost, and the summation of the parts cost, the total cost is calculated. A printout is then made so that the user can follow the steps that the computer makes. Following this, some method must be employed to determine if cost is a minimum. If it is, then a final printout can be made. If it is not, then an option is shown as to how one wants to optimize. This can be accomplished by the user reading in another set of part numbers or the computer automatically can select a set in the manner described in Section 3.5 using the search matrix library. This procedure is repeated in an iterative manner until the optimum design is reached.

There are many associated details that are not shown in Figure 3.1. This diagram, however, gives the general outline of the procedure.

### 3.2 GENERATION OF OBJECT FUNCTIONS

The first question that must be answered in an optimization problem is, "What is to be optimized and what is optimum?" Often, this is not a trivial problem in itself since there are many separate and usually conflicting factors; i.e., minimum cost, maximum accuracy, small volume, best response, etc. These factors may be considered simultaneously by defining a scalar $F$ of the form

$$
\begin{equation*}
F=\sum_{i=1}^{k} A_{i}\left(Y_{i}-D_{i}\right)^{2} \tag{3.1}
\end{equation*}
$$

where:

$$
\begin{aligned}
\mathrm{F}= & \text { object function to be minimized } \\
\mathrm{k}= & \text { number of desired properties } \\
\mathrm{A}_{\mathbf{i}}= & \text { weight factor selected to give the } i^{\text {th }} \\
& \text { property the desired priority } \\
\mathbf{Y}_{\mathbf{i}}= & \text { current value of } i^{\text {th }} \text { property } \\
\mathrm{D}_{\mathbf{i}}= & \text { desired value for } i^{\text {th }} \text { property }
\end{aligned}
$$

A serious difficulty inherent in this approach, however, consists in finding a set of weighting factors $A_{1}, A_{2}, \ldots, A_{k}$ such that scaling between the various terms is properly considered in order to maintain sensitivity and obtain good convergence. Considering

$$
1=
$$

properties such as accuracy, weight, cost and response, these weight selections often become subjective in nature.

It is proposed in this thesis that an entirely different object function shall be used. It is founded on the competitive philosophy that the manufacturer wants a design that fulfills the customer requirements at minimum overall cost. With this result, he can either maximize his chances of competing or if his sale price is "fixed" he maximizes his profits. Using this minimum cost philosophy, an appropriate object function can be generated in the following manner.

The total cost to build a given number of systems is represented by the equation

$$
\begin{align*}
& \text { Total }  \tag{3.2}\\
& \text { Cost }
\end{aligned}=\begin{aligned}
& \text { Number } \\
& \text { Built }
\end{align*}\left[\begin{array}{l}
\text { Labor } \\
\text { Cost }
\end{array}+\sum \begin{array}{l}
\text { Component } \\
\text { Costs }
\end{array}\right]\left[\begin{array}{l}
\text { Overhead } \\
\text { Ratio }
\end{array}\right]
$$

However, the number that must be built for a given contract is given by

$$
\frac{\text { Number }}{\text { Built }}=\frac{\text { Number } \text { Required }}{\left[\begin{array}{ll}
\text { Rejection }  \tag{3.3}\\
\text { Ratio }
\end{array}\right]}
$$

Thus, we have for the total cost


Since the number of required units and (1+overhead ratio) are product terms which are not functions of the components, one obtains the same cost minimizing set of components using the function

$$
\text { Cost }=\frac{\begin{array}{l}
\text { Labor }  \tag{3.5}\\
\text { Cost }
\end{array}+\sum \begin{array}{l}
\text { Component } \\
\text { Costs }
\end{array}}{1-\begin{array}{l}
\text { Rejection } \\
\text { Ratio }
\end{array}}
$$

Equation (3.5) is the object function used in this thesis for what is defined later as "the fine search mode." When it is at a minimum, the desired optimum set of components has been defined. However, one problem may exist in the early portion of the iteration cycle. That is, the design can be so far away from specification that, for all practical purposes, the rejection ratio is unity, the denominator of (3.5) goes to zero, resulting in infinite cost. As long as this occurs, (3.5) has no practical value. In fact, one loses all sensitivity in calculating partials, and there is no way of telling if one design is better than another. For this reason, a "coarse search mode" is defined. Its corresponding object function is:

$$
\begin{equation*}
F=\sum_{i=1}^{k} A_{i} R_{i}\left(Y_{i}-S_{i}\right)^{2} \tag{3.6}
\end{equation*}
$$

where
$\mathrm{F}=$ object function to be minimized
$k=$ number of specifications to be met
$A_{i}=$ weight factor for $i^{\text {th }}$ specification
$R_{i}=$ rejection ratio for $i^{\text {th }}$ specification


$$
\begin{aligned}
Y_{i}= & \text { calculated system performance } 3 \text { sigma } \\
& \text { limit corresponding to } i^{\text {th }} \text { specification } \\
S_{i}= & i^{\text {th }} \text { specification limit }
\end{aligned}
$$

It should be further noted that

$$
\begin{aligned}
& Y_{i}=\bar{Y}_{i}-3 \sigma_{Y_{i}} \text { if } S_{i} \text { is a lower limit, and } \\
& Y_{i}=\bar{Y}_{i}+3 \sigma_{Y_{i}} \text { if } S_{i} \text { is an upper limit. }
\end{aligned}
$$

Since Equation (3.6) is used only in the coarse search mode, selection of the weight factors is not too critical. For this study, $A_{i}$ was set at $1 / S_{i}{ }^{2}$ except for the case when $S_{i}$ equals zero and then $A_{i}$ was arbitrarily set equal to unity.

In the coarse search mode, cost is neglected in an attempt to determine the performance such that the rejection ratio becomes less than unity. The incorporation of the $R_{i}$ term in (3.6) greatly aids in the accomplishment of this condition. First it nulls each term in the summation which represents an overdesigned condition (i.e. $\mathrm{R}_{\mathrm{i}}=0$ ) and secondly it applies a. linearily increasing weight on the others according to their significance.

Once each of the $R_{i}$ 's is driven less than unity, the cost becomes finite, and the optimization process is switched from the coarse to the fine search where (3.5) is used as the object function.

### 3.3 CALCULATION OF REJECTION RATIO

Let $s_{1}, \cdots, s_{k}$ be the $k$ specification limits for a given design, e.g., static accuracy, overshoot, etc. There corresponds
$1$
then, $k$ random variables $Y_{1}, \cdots, Y_{k}$ that represent the actual performance to be expected. Since these are a function of the $n$ component equations, one can write as before that

$$
\left[\begin{array}{c}
Y_{1}  \tag{3.7}\\
Y_{2} \\
\vdots \\
Y_{k}
\end{array}\right]=\left[\begin{array}{l}
F_{1}\left(x_{1}, \cdots, x_{n}\right) \\
F_{2}\left(x_{1}, \cdots, x_{n}\right) \\
\vdots \\
F_{k}\left(x_{1}, \cdots, x_{n}\right)
\end{array}\right]
$$

Looking at small perturbations

$$
\left[\begin{array}{c}
\Delta Y_{1}  \tag{3.8}\\
\Delta Y_{2} \\
\vdots \\
\Delta Y_{k}
\end{array}\right]=\left[\begin{array}{cccc}
a_{11} & a_{12} & \cdots & a_{1 n} \\
& & & \\
a_{21} & a_{22} & \cdots & a_{2 n} \\
\vdots & \vdots & & \vdots \\
a_{k 1} & a_{k 2} & \cdots & a_{k n}
\end{array}\right] \quad\left[\begin{array}{c}
\Delta X_{1} \\
\Delta X_{2} \\
\vdots \\
\Delta X_{n}
\end{array}\right]
$$

where the $k \times n$ matrix has the general element $a_{i j}=\frac{\partial Y_{i}}{\partial X_{j}}$ and it therefore is identical to the Jacobian matrix ( J ) as defined in (2.23).

The joint density of the Y's is given by:

$$
\begin{equation*}
f_{Y_{1}}, Y_{2}, \cdots{ }_{Y_{k}}\left(y_{1}, y_{2}, \cdots y_{k}\right)=\frac{\left.e^{-\left[(Y-\bar{Y}) M_{Y}\right.}{ }^{-1}(Y-\bar{Y})^{T}\right]}{(2 \pi)^{k / 2} \sqrt{\left|M_{Y}\right|}} \tag{3.9}
\end{equation*}
$$

where:

$$
(Y-\bar{Y})=\left[\left(y_{1}-\bar{y}_{1}\right),\left(y_{2}-\bar{y}_{2}\right), \cdots,\left(y_{k}-\bar{y}_{k}\right)\right]
$$

and the $(k \times k)$ covariance matrix $M_{Y}$ is

$$
\begin{equation*}
M_{Y}=J M_{X} J^{T} \tag{3.10}
\end{equation*}
$$

Since the component performance parameters are assumed independent and $\sigma_{X_{i}}=0$ if $X_{i}$ is a component cost, one can write the component covariance matrix $M_{X}$ as

$$
M_{x}=\left[\begin{array}{cccc}
\sigma_{x_{1}}{ }^{2} & 0 & \cdots & 0  \tag{3.11}\\
0 & \sigma_{x_{2}}{ }^{2} & \cdots & 0 \\
\vdots & \vdots & & \vdots \\
0 & 0 & \cdots & \sigma_{x_{n}}{ }^{2}
\end{array}\right]
$$

Since the total rejection ratio $R$ is the probability of a design falling outside of the specification, and assuming that the specification limits are constant, it is given by
$R=1-\int_{L_{11}}^{L_{12}} \int_{L_{21}}^{L_{22}} \cdots \int_{L_{k 1}}^{L_{k 2}} f_{Y_{1}}, Y_{2}, \cdots,{ }_{Y_{k}}\left(y_{1}, y_{2}, \cdots, y_{k}\right) d y_{1} d y_{2} \cdots d y_{k}$
where:

$$
\begin{aligned}
& \left\{\begin{array}{l}
L_{i 1}=-\infty \\
L_{i 2}=S_{i}
\end{array}\right\} \quad \text { for the } i^{\text {th }} \text { specification an upper bound } \\
& \left\{\begin{array}{ll}
L_{i 1}=S_{i} \\
L_{i 2}=\infty
\end{array}\right\} \quad \text { for the } i^{\text {th }} \text { specification a lower bound }
\end{aligned}
$$

and

$$
f_{Y_{1}}, Y_{2}, \cdots, Y_{k}\left(y_{1}, y_{2}, \cdots, y_{k}\right) \text { is given by (3.9). }
$$

In order to evaluate $R$ using (3.12), one must evaluate the multiple integral of dimension $k$ where $k=8$ for the example in this study. This can be accomplished using numerical techniques [28], and [29], however, the process is very time consuming. In the interest of minimizing computer time, three alternate procedures are considered.

First one could use the upper bound on $R$ which is simply

$$
\begin{align*}
R(\text { upper bound }) & =\sum_{i=1}^{k} R_{i} \quad \text { if } \sum_{i=1}^{k} R_{i}<1 \\
& =1 \quad \text { otherwise } \tag{3.13}
\end{align*}
$$

where $R_{i}$ is the individual rejection ratio corresponding to the $i^{\text {th }}$ specification and is calculated as

$$
\begin{equation*}
R_{i}=1-\frac{1}{\sqrt{2 \pi \sigma_{Y_{i}^{2}}}} \int_{L_{i 1}}^{L_{i 2}} e^{-\frac{1}{2}\left(\frac{{ }^{y-\mu_{Y_{i}}}}{\sigma_{Y_{i}}}\right)^{2}} d y \tag{3.14}
\end{equation*}
$$

where the limits of the integral are as defined for (3.12).
Equation (3.14) can be evaluated by using the standard error function

$$
\begin{equation*}
\operatorname{ERF}(z)=\frac{2}{\sqrt{\pi}} \int_{0}^{z} e^{-u^{2}} d u \tag{3.15}
\end{equation*}
$$

for the upper limit case

$$
\begin{align*}
R_{i} & =0.5\left[1-\operatorname{ERF}\left(\frac{S_{i}-\mu_{Y_{i}}}{\sqrt{2} \sigma_{Y_{i}}}\right)\right] \quad \text { for } S_{i} \geq{ }^{\mu_{Y_{i}}} \\
& =0.5\left[1+\operatorname{ERF}\left(\frac{{ }^{\mu} Y_{i}{ }^{-S_{i}}}{\sqrt{2} \sigma_{Y_{i}}}\right)\right] \quad \text { for } \quad S_{i}<{ }^{\mu_{Y_{i}}} \tag{3.16}
\end{align*}
$$

and for the lower limit case

$$
\begin{align*}
R_{i} & =0.5\left[1+\operatorname{ERF}\left(\frac{s_{i}-\mu_{Y_{i}}}{\sqrt{2} \sigma_{Y_{i}}}\right)\right] \quad \text { for } \quad S_{i} \geq{ }_{Y_{Y_{i}}} \\
& =0.5\left[1-\operatorname{ERF}\left(\frac{{ }_{Y_{i}}-S_{i}}{\sqrt{2} \sigma_{Y_{i}}}\right)\right] \quad \text { for } \quad S_{i}<\mu_{Y_{i}} \tag{3.17}
\end{align*}
$$

A second possibility for approximating the total rejection ratio $R$ is to use the lower bound given by

$$
\begin{equation*}
R(\text { lower bound })=R_{j} \tag{3.18}
\end{equation*}
$$

where:

$$
R_{j} \leq R_{i} \quad \text { for all } \quad 1 \leq i \leq k
$$

Since (3.13) represents an overdesigned case and (3.18) an underdesigned case, it would be wise to have available an approximation that lies between these extremes. A quantity which has this property is

$$
\begin{equation*}
R(\text { independent })=1-\prod_{i=1}^{k}\left(1-R_{i}\right) \tag{3.19}
\end{equation*}
$$

which is equal to the true $R$ for the case when the $Y^{\prime} s$ are independent.

For the example program, the user is given the opportunity of selecting either the $R$ (upper bound) or $R$ (independent) approximations. The $R$ (lower bound), although readily available, is eliminated as a choice since it is never on the safe side.

One difficulty remains since (3.16) cannot be used to calculate the rejection ratio for the null oscillation specification. This specification that no null oscillation shall exist is converted by the computer to a specification limit on the actual backlash. This limit is not a constant but a random variable computed as described in Appendix A. Therefore, the rejection ratio must be computed by examing two frequency distributions, namely that of the allowable backlash and that of the actual backlash of the geartrain being considered. Thus a separate subroutine was written to calculate $R_{6}$ the derivation of which is explained in the remainder of this section

For this derivation only, the random variable $Y$ is used to represent the actual backlash and $S$ the allowable backlash. Since both are assumed to be normally distributed their density functions are defined as

$$
\begin{align*}
& g_{1}(y)=\frac{1}{\sqrt{2 \pi \sigma_{Y}^{2}}} e^{-\frac{1}{2}\left(\frac{y-\mu_{Y}}{\sigma_{Y}}\right)^{2}}  \tag{3.20}\\
& g_{2}(s)=\frac{1}{\sqrt{2 \pi \sigma_{S}^{2}}} e^{-\frac{1}{2}\left(\frac{s-\mu_{S}}{\sigma_{S}}\right)^{2}}
\end{align*}
$$

and the corresponding rejection ratio is given by the probability that $Y>S$ as

$$
\begin{equation*}
R_{6}=P(Y>S)=\int_{-\infty}^{\infty} \int_{-\infty}^{y} g(y, s) d s d y \tag{3.22}
\end{equation*}
$$

and since $Y$ and $S$ are independent

$$
\begin{equation*}
g(y, s)=g_{1}(y) g_{2}(s) \tag{3.23}
\end{equation*}
$$

By using (3.23 in (3.22) and substituting in for $g_{1}(y)$ and $g_{2}(s)$ using (3.20) and (3.21), and simplifying, (3.22) can be written as

$$
\begin{equation*}
R_{6}=\frac{1}{\sqrt{2 \pi \sigma_{Y}^{2}}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}\left(\frac{y-\mu_{Y}}{\sigma_{Y}}\right)^{2}} \frac{1}{\sqrt{2 \pi \sigma_{S}^{2}}} \int_{-\infty}^{y} e^{-\frac{1}{2}\left(\frac{s-\mu_{S}}{\sigma_{S}}\right)^{2} d s d y} \tag{3.24}
\end{equation*}
$$

## Letting

$$
\begin{equation*}
A(y)=\frac{1}{\sqrt{2 \pi \sigma_{S}^{2}}} \int_{-\infty}^{y} e^{-\frac{1}{2}\left(\frac{s-\mu_{S}}{\sigma_{S}}\right)^{2}} d s \tag{3.25}
\end{equation*}
$$

where $A(y)$ can be evaluated by using the standard error function as before, one obtains

$$
\begin{align*}
A(y) & =0.5\left[1+\operatorname{ERF}\left(\frac{y-\mu_{S}}{\sqrt{2} \sigma_{S}}\right)\right] \quad \text { for } y \geq \mu_{S} \\
& =0.5\left[1-\operatorname{ERF}\left(\frac{\mu_{S}-y}{\sqrt{2} \sigma_{S}}\right)\right] \quad \text { for } y<\mu_{S} \tag{3.26}
\end{align*}
$$

and since one is interested only in the region inside the 3 -sigma limits, $R_{6}$ is evaluated as

$$
\begin{equation*}
R_{6}=\frac{1}{\sqrt{2 \pi \sigma_{Y}{ }^{2}}} \int_{\mathrm{YMIN}}^{Y M A X} A(y) e^{-\frac{\left(y-\mu_{Y}\right)^{2}}{2 \sigma_{Y}^{2}}} d y \tag{3.27}
\end{equation*}
$$

where $A(y)$ is evaluated using (3.26) and YMAX and YMIN are taken to be $\mu_{Y}+3 \sigma_{Y}$ and $\mu_{Y}-3 \sigma_{Y}$ respectively.

### 3.4 OBJECT FUNCTION DERIVATIVES

It is of necessity that the partial derivatives of the object function be calculated in the steepest ascent method of optimization. If these derivatives were somehow known for the direct search technique, it would be of advantage since one could then conduct exploratory moves in descending order of importance. In our case, it would be a major task to perturbate each of the component parameters again and calculate the resulting change in the object function to obtain the partial derivatives. It is shown, however, that these can be obtained directly from the Jacobian matrix which is already available from the tolerance calculations, namely, Equation (2.23). This is accomplished in the following manner as derived first for the fine search and then for the coarse search.

The object function used in fine search, Equation (3.5), can be written as

$$
\begin{equation*}
C(X)=[K+f(X)][1-R(X)]^{-1} \tag{3.28}
\end{equation*}
$$

where

$$
\begin{array}{ll}
X & =\text { component parameter vector }\left[X_{1}, X_{2}, \cdots, X_{n}\right] \\
C(X) & =\text { total system cost } \\
K & =\text { labor cost } \\
R(X) & =\text { rejection ratio } \\
f(X) & =\Sigma \text { component cost }
\end{array}
$$

Taking the partial derivative of $C$ with respect to $X_{i}$

$$
\begin{equation*}
\frac{\partial C}{\partial X_{i}}=\frac{\partial f(X)}{\partial X_{i}}(1-R(X))^{-1}+(K+f(X))(1-R(X))^{-2} \frac{\partial R}{\partial X_{i}} \tag{3.29}
\end{equation*}
$$

and expanding to include all $\mathrm{X}_{\mathrm{i}}$

$$
\begin{align*}
& {\left[\frac{\partial C}{\partial X_{1}}, \frac{\partial C}{\partial X_{2}}, \cdots, \frac{\partial C}{\partial X_{n}}\right]=\frac{1}{1-R(X)}\left[\frac{\partial f}{\partial X_{1}}, \frac{\partial f}{\partial X_{2}}, \cdots, \frac{\partial f}{\partial X_{n}}\right]} \\
& \quad+\frac{K+f(X)}{(1-R(X))^{2}}\left[\frac{\partial R}{\partial X_{1}}, \frac{\partial R}{\partial X_{2}}, \cdots, \frac{\partial R}{\partial X_{n}}\right] \tag{3.30}
\end{align*}
$$

The latter vector ( $\partial R / \partial X$ ) can be obtained by making use of the Jacobian defined by (2.23). Thus

$$
\left[\frac{\partial R}{\partial X_{1}}, \frac{\partial R}{\partial X_{2}}, \cdots, \frac{\partial R}{\partial X_{n}}\right]=\left[\frac{\partial R}{\partial Y_{1}}, \frac{\partial R}{\partial Y_{2}}, \cdots, \frac{\partial R}{\partial Y_{k}}\right]\left[\begin{array}{cc}
\frac{\partial Y_{1}}{\partial X_{1}} \cdots & \frac{\partial Y_{1}}{\partial X_{n}}  \tag{3.31}\\
\vdots & \vdots \\
\frac{\partial Y_{k}}{\partial X_{1}} \cdots & \frac{\partial Y_{k}}{\partial X_{n}}
\end{array}\right]
$$

Substituting (3.31) into (3.30) one obtains the desired matrix equation for the fine search cost derivative vector as

$$
\begin{align*}
& {\left[\frac{\partial C}{\partial X_{1}}, \frac{\partial C}{\partial X_{2}}, \cdots, \frac{\partial C}{\partial X_{n}}\right]=\frac{1}{1-R(X)}\left[\frac{\partial f}{\partial X_{1}}, \frac{\partial f}{\partial X_{2}}, \cdots, \frac{\partial f}{\partial X_{k}}\right]} \\
& \quad+\frac{K+f(X)}{(1-R(X))^{2}}\left[\frac{\partial R}{\partial Y_{1}}, \frac{\partial R}{\partial Y_{2}}, \cdots, \frac{\partial R}{\partial Y_{k}}\right]\left[\begin{array}{cc}
\frac{\partial Y_{1}}{\partial X_{1}} \cdots \frac{\partial Y_{1}}{\partial X_{n}} \\
\vdots \\
\vdots \\
\frac{\partial Y_{k}}{\partial X_{1}} \cdots \frac{\partial Y_{k}}{\partial X_{n}}
\end{array}\right] \tag{3.32}
\end{align*}
$$

where:

$$
\frac{\partial f}{\partial X_{i}}= \begin{cases}1 & \text { if } X_{i} \text { is a component cost }  \tag{3.33}\\ 0 & \text { otherwise }\end{cases}
$$

and the vector

$$
\frac{\partial R}{\partial Y_{1}}, \frac{\partial R}{\partial Y_{2}}, \cdots, \frac{\partial R}{Y_{k}}
$$

is referred to as the "rejection ratio derivative vector" and given the notation $\partial R / \partial Y$.

The calculation of the $\partial R / \partial Y$ vector, as required for the fine search mode, depends on the particular equation used in approximating the rejection ratio $R$ [i.e., (3.13) or (3.19)]. Consider first the case where $R$ is approximated by the upper bound. Since in the fine search mode

$$
\sum_{i=1}^{k} R_{i}<1
$$

one has

$$
\begin{equation*}
R(\text { upper bound })=R_{1}+R_{2}+\cdots+R_{k} \tag{3.34}
\end{equation*}
$$

and since $R_{j}$ is a function of $Y_{i}$ only for $i=j$

$$
\begin{equation*}
\frac{\partial R \text { (upper bound) }}{\partial Y_{i}}=\frac{\partial R_{i}}{\partial Y_{i}} \text { for } i=1,2, \ldots, k \tag{3.35}
\end{equation*}
$$

and only the partials of the individual rejection ratios are required. This is also shown to be the case when $R$ is approximated by using the case where the Y 's are assumed independent as given by

$$
\begin{equation*}
R(\text { independent })=1-\left(1-R_{1}\right)\left(1-R_{2}\right) \cdots\left(1-R_{k}\right) \tag{3.36}
\end{equation*}
$$

Again using the fact that $R_{i}$ is only a function of $Y_{i}$, one obtains

$$
\begin{equation*}
\frac{\partial R(\text { independent })}{\partial Y_{i}}=\frac{\partial R_{i}}{\partial Y_{i}} \prod_{\substack{j=1 \\ j \neq i}}^{k}\left(1-R_{j}\right) \tag{3.37}
\end{equation*}
$$

The task remaining, then, is to obtain expressions for $\partial R_{i} / \partial Y_{i}$.

For the case where the specification limit is a constant, the magnitude of $\partial R_{i} / \partial Y_{i}$ is given by the $Y_{i}$ density function evaluated at the point $y_{i}=S_{i}$ and the sign of $\partial R_{i} / \partial Y_{i}$ depends on whether $S_{i}$ is an upper or a lower bound. That is

$$
\begin{equation*}
\frac{\partial R_{i}}{\partial Y_{i}}=\frac{ \pm 1}{\sqrt{2 \pi \sigma_{Y}^{2}}} e^{-\frac{1}{2}\left[\frac{S_{i}-\mu_{Y_{i}}}{\sigma_{Y_{i}}}\right]^{2}} \tag{3.38}
\end{equation*}
$$

where

$$
\begin{aligned}
S_{i} & =i^{\text {th }} \text { specification limit } \\
{ }^{\mu_{Y_{i}}} & =\text { mean value of } Y_{i} \text { distribution } \\
\sigma_{Y_{i}} & =\text { standard deviation of } Y_{i} \text { distribution }
\end{aligned}
$$

and the + sign is taken if $S_{i}$ is an upper limit and the - sign is taken if $S_{i}$ is a lower limit.

For the case where the specification limit is not a constant but a random variable (e.g., $\mathrm{S}_{6}$ ), the corresponding equation is

$$
\begin{equation*}
\frac{\partial R_{i}}{\partial Y_{i}}=\int_{-\infty}^{\infty} g_{1}(z) g_{2}(z) d z \tag{3.39}
\end{equation*}
$$

where the density functions $g_{1}(\cdot)$ and $g_{2}(\cdot)$ are defined by (3.20) and (3.21) respectively. The solution to (3.39) is approximated in the example design program by using numerical integration over the region from $\mu_{Y_{i}}{ }^{-3 \sigma_{Y_{i}}}$ to ${ }_{\mu_{Y_{i}}}{ }^{+3 \sigma_{Y_{i}}}$.

In summary, Equation (3.32) gives the required partial derivatives of cost with respect to each component parameter. The necessary elements of the rejection ratio derivative vector are obtained using either (3.35) or (3.37) and with the $\partial R_{i} / \partial Y_{i}$ entries furnished by either (3.38) or (3.39) as the requirements dictate. A similar development is presented now for the coarse search mode.

The object function used for coarse search is of the form [see (3.6)]

$$
F(X)=A_{1} R_{1}(X)\left[Y_{1}(X)-S_{1}\right]^{2}+A_{2} R_{2}(X)\left[Y_{2}(X)-S_{2}\right]^{2}+\cdots+A_{k} R_{k}(X)\left[Y_{k}(X)-S_{k}\right]^{2}
$$

where for the case of the example program the value used for $S_{6}$ is taken to be its mean value.

Taking the partial derivative of $F$ with respect to $X_{1}$

$$
\begin{align*}
\frac{\partial F}{\partial X_{1}} & =2 A_{1} R_{1}\left(Y_{1}-S_{1}\right) \frac{\partial Y_{1}}{\partial X_{1}}+A_{1}\left(Y_{1}-S_{1}\right)^{2} \frac{\partial R_{1}}{\partial X_{1}} \\
& +2 A_{2} R_{2}\left(Y_{2}-S_{2}\right) \frac{\partial Y_{2}}{\partial X_{1}}+A_{2}\left(Y_{2}-S_{2}\right)^{2} \frac{\partial R_{2}}{\partial X_{1}} \\
& +\cdots+2 A_{k} R_{k}\left(Y_{k}-S_{k}\right) \frac{\partial Y_{k}}{\partial X_{1}}+A_{k}\left(Y_{k}-S_{k}\right)^{2} \frac{\partial R_{k}}{\partial X_{1}} \tag{3.41}
\end{align*}
$$

Thus, in total vector form:

$$
\begin{align*}
& {\left[\frac{\partial F}{\partial X_{1}}, \frac{\partial F}{\partial X_{2}}, \cdots, \frac{\partial F}{\partial X_{n}}\right]=2\left[\begin{array}{c}
A_{1} R_{1}\left(Y_{1}-S_{1}\right) \\
A_{2} R_{2}\left(Y_{2}-S_{2}\right) \\
\vdots \\
A_{k} R_{k}\left(Y_{k}-S_{k}\right)
\end{array}\right]\left[\begin{array}{ccc}
\frac{\partial Y_{1}}{\partial X_{1}} & \frac{\partial Y_{1}}{\partial X_{2}} \cdots & \frac{\partial Y_{1}}{\partial X_{n}} \\
\frac{\partial Y_{2}}{\partial X_{1}} & \frac{\partial Y_{2}}{\partial X_{2}} \cdots & \frac{\partial Y_{2}}{\partial X_{n}} \\
\vdots & \vdots & \vdots \\
\frac{\partial Y_{k}}{\partial X_{1}} & \frac{\partial Y_{k}}{\partial X_{2}} & \frac{\partial Y_{k}}{\partial X_{n}}
\end{array}\right]} \\
& +\left[\begin{array}{c}
A_{1}\left(Y_{1}-S_{1}\right)^{2} \\
A_{2}\left(Y_{2}-S_{2}\right)^{2} \\
\vdots \\
A_{k}\left(Y_{k}-S_{k}\right)^{2}
\end{array}\right]^{T}\left[\begin{array}{ccc}
\frac{\partial R_{1}}{\partial X_{1}} & \frac{\partial R_{1}}{\partial X_{2}} & \frac{\partial R_{1}}{\partial X_{n}} \\
\frac{\partial R_{2}}{\partial X_{1}} & \frac{\partial R_{2}}{\partial X_{2}} & \frac{\partial R_{2}}{\partial X_{n}} \\
\vdots & \vdots & \vdots \\
\frac{\partial R_{k}}{\partial X_{1}} & \frac{\partial R_{k}}{\partial X_{2}} & \cdots \\
\frac{\partial R_{k}}{\partial X_{n}}
\end{array}\right] \tag{3.42}
\end{align*}
$$

Using the further relationship that:

$$
\left[\begin{array}{cccc}
\frac{\partial R_{1}}{\partial X_{1}} & \frac{\partial R_{1}}{\partial X_{2}} & \cdots & \frac{\partial R_{1}}{\partial X_{n}}  \tag{3.43}\\
\frac{\partial R_{2}}{\partial X_{1}} & \frac{\partial R_{2}}{\partial X_{2}} & \cdots & \frac{\partial R_{2}}{\partial X_{n}} \\
\vdots & \vdots & & \vdots \\
\frac{\partial R_{k}}{\partial X_{1}} & \frac{\partial R_{k}}{\partial X_{2}} & \cdots & \frac{\partial R_{k}}{\partial X_{n}}
\end{array}\right]=\left[\begin{array}{cccc}
\frac{\partial R_{1}}{\partial Y_{1}} & \frac{\partial R_{1}}{\partial Y_{2}} \cdots & \frac{\partial R_{1}}{\partial Y_{k}} \\
\frac{\partial R_{2}}{\partial Y_{1}} & \frac{\partial R_{2}}{\partial Y_{2}} \cdots & \frac{\partial R_{2}}{\partial Y_{k}} \\
\vdots & \vdots & & \vdots \\
\frac{\partial R_{k}}{\partial Y_{1}} & \frac{\partial R_{k}}{\partial Y_{2}} \cdots & \frac{\partial R_{k}}{\partial Y_{k}}
\end{array}\right]\left[\begin{array}{cccc}
\frac{\partial Y_{1}}{\partial X_{1}} & \frac{\partial X_{2}}{\partial X_{2}} & \cdots & \frac{\partial Y_{n}}{\partial X_{n}} \\
\frac{\partial Y_{2}}{\partial X_{1}} & \frac{\partial Y_{2}}{\partial X_{2}} \cdots & \frac{\partial Y_{2}}{\partial X_{n}} \\
\vdots & \vdots & \vdots \\
\frac{\partial Y_{k}}{\partial X_{1}} & \frac{\partial Y_{k}}{\partial X_{2}} \cdots & \frac{\partial Y_{k}}{\partial X_{n}}
\end{array}\right]
$$

and noting that $\frac{\partial R_{i}}{\partial Y_{j}}=0$ for all $i \neq j$, and substituting (3.43) into (3.42), one obtains

$$
\left[\frac{\partial F}{\partial X_{1}}, \frac{\partial F}{\partial X_{2}}, \cdots, \frac{\partial F}{\partial X_{n}}\right]=2\left[\begin{array}{c}
A_{1}\left(Y_{1}-S_{1}\right) R_{1}+\left(Y_{1}-S_{1}\right)^{2} \frac{\partial R_{1}}{\partial Y_{1}} \\
A_{2}\left(Y_{2}-S_{2}\right) R_{2}+\left(Y_{2}-S_{2}\right)^{2} \frac{\partial R_{2}}{\partial Y_{2}} \\
\vdots \\
A_{k}\left(Y_{k}-S_{k}\right) R_{k}+\left(Y_{k}-S_{k}\right)^{2} \frac{\partial R_{k}}{\partial Y_{k}}
\end{array}\right]^{T}\left[\begin{array}{ccc}
\frac{\partial Y_{1}}{\partial X_{1}} & \frac{\partial Y_{1}}{\partial X_{2}} \cdots \frac{\partial Y_{1}}{\partial X_{n}} \\
\frac{\partial Y_{2}}{\partial X_{1}} & \frac{\partial Y_{2}}{\partial X_{2}} \cdots \frac{\partial Y_{2}}{\partial X_{n}} \\
\vdots & \vdots & \vdots \\
\frac{\partial Y_{k}}{\partial X_{1}} & \frac{\partial Y_{k}}{\partial X_{2}} \cdots \frac{\partial Y_{k}}{\partial X_{n}}
\end{array}\right]
$$

Equation (3.44) gives the desired partial derivatives of the coarse search object function with respect to each component parameter in the system. Again, like (3.32), it is in terms of the already available Jacobian matrix and no further parameter perturbations are required.

### 3.5 DESIGN PROGRAM STRATEGY

The design program developed as part of this study has two basic operating options - analysis and directed search. When operating with the analysis option, the four component part numbers required for each analysis may be either read in from cards or selected at random by the program. In either case, as many consecutive runs are made as requested and a final printout is provided summarizing the best design obtained. Thus the engineer can make a rapid evaluation of a selected number of designs of his choosing, or, he can perform Monte Carlo runs by letting the computer select the part numbers at random.

With the directed search option, the computer program uses the object derivatives in connection with search matrices to direct the next component selection in an attempt to reduce the object function. This process is repeated in an iterative fashion until a local minimum is obtained. Since there is no guarantee that this condition is the absolute minimum, numerous starting points are employed and the one with the lowest cost is assumed to be the best design. The starting points for each search may be specified by the user or otherwise selected at random by the program.

The generation of the search matrices is a prerequisite to a directed search. A separate search matrix is used along with each component library and their generation automatically follows each library update. These matrices consist of an ordered array of the component part numbers defined by

$$
S_{i}=\left[\begin{array}{cccc}
s_{11} & s_{12} & \cdots & s_{1 \ell}  \tag{3.45}\\
s_{21} & s_{22} & \cdots & s_{2 \ell} \\
\vdots & \vdots & & \vdots \\
s_{m 1} & s_{m 2} & \cdots & s_{m \ell}
\end{array}\right]
$$

where
$\begin{aligned} \ell= & \text { the number of parameters used to describe the } \\ & i^{\text {th }} \text { component }\end{aligned}$
$m=$ the number of part numbers for $i^{\text {th }}$ component stored in the library
$s_{n j}=a$ component part number for $1 \leq n \leq m$ and $1 \geq j \geq \ell$

Each column of $S_{i}$ corresponds to a particular parameter of the $i^{\text {th }}$ component and the entries of that column consist of all the $i^{\text {th }}$ component part numbers arranged in ascending order of the mean value of that parameter. That is, let the $j^{\text {th }}$ column of $S_{i}$ correspond to the $k^{\text {th }}$ component parameter of the $X$ vector. Then $s_{1 j}$, $s_{2 j}, \cdots, s_{m j}$ are chosen such that

$$
\begin{equation*}
\overline{X_{k}\left(s_{1 j}\right)} \leq \overline{X_{k}\left(s_{2 j}\right)} \leq \overline{X_{k}\left(s_{3 j}\right)} \leq \cdots \leq \overline{X_{k}\left(s_{m j}\right)} \tag{3.46}
\end{equation*}
$$

where

$$
\begin{aligned}
\overline{X_{k}\left(S_{n} j\right)} & \text { signifies the mean value of the component parameter } \\
& X_{k} \text { for the part number stored in location } s_{n j}
\end{aligned}
$$

In order to explain the strategy used by the design program to conduct a search, the following definitions are established:


Thus a search is made up of many sub-searches and each of the latter are in turn made up of numerous exploratory moves. Each exploratory move consists of changing one component part number while keeping the others fixed at the base point. Once an exploratory move
results in "success," the move is defined as a new base point and a new sub-search is started. This process is repeated until all the exploratory moves of a sub-search are exhausted and no success is found. The base point for this last sub-search defines the local minimum.

The following ten steps describe the general pattern of the program's search strategy:

1) The object function being minimized is SCALAR [defined as $F$ in Equation (3.6)] while in the coarse search mode and COST [Equation (3.5)] while in the fine search mode. The program is in the coarse search mode as long as the total rejection ratio [Equation (3.13) or (3.19)] is equal to unity, once less than unity the program switches to the fine search mode.
2) Each time a lower object function is found, the corresponding part numbers are stored as a new base point.
3) At each new base point, calculations are made to establish the object function derivative vector using Equation (3.44) for the coarse search mode and (3.32) for the fine search mode.
4) Priority and direction vectors are established as the bases for making exploratory moves. The priority vector (IPAR) consists of a re-ordering of the component parameter numbers (i.e., subscripts of the $X$ parameter vector) such that

$$
\begin{equation*}
\left|\frac{\partial \text { object }}{\partial X_{\text {IPAR }_{1}}}\right| \geq\left|\frac{\partial \text { object }}{\partial X_{I P A R_{2}}}\right| \geq \cdots \geq\left|\frac{\partial \text { object }}{\partial X_{\text {IPAR }}^{m}}\right| \tag{3.47}
\end{equation*}
$$

where $m$, the dimension of IPAR, equals the number of component parameters excluding the load (e.g., with the vector defined in Table 2.5, $m=21$ ). The direction vector (IDEX) is defined by

Thus

$$
\begin{aligned}
\text { IDEX }_{\text {II }} & =+1 \text { if the } I^{\text {th }} \text { parameter should be increased } \\
& =-1 \text { if the } I^{\text {th }} \text { parameter should be decreased }
\end{aligned}
$$

in order to achieve a reduction in the object function.
5) A "sub-search progress number," denoted by the symbol II, is used by the program as the subscript for the IPAR and IDEX vectors. It is initialized equal to unity (i.e., II = l) at the beginning of each sub-search and incremented under program control as the sub-search progresses. As II is increased from 1 to $m$, IPAR $_{\text {II }}$ corresponds to the component parameter numbers having decreasing sensitivity values with respect to the object function. Likewise, IDEX $_{\text {IPAR }_{\text {II }}}$ corresponds to the desired direction the IPAR $_{\text {II }}$ parameter is to be changed.
6) Each exploratory move is initiated by calling a subroutine, named SEARCH, to select the new part number which is to be investigated. This is accomplished using the statement:

$$
\text { CALL SEARCH[IDEX } \left.{ }_{\text {IPAR }}^{\text {II }}, \text { IPAR }_{\text {II }}, I \mathrm{IPN}_{J J J}, I B O U N D\right]
$$

where

and for this example, the JJJ subscript is established from Table 2.5 as

$$
\begin{aligned}
\mathrm{JJ} & =1 \text { for } 1 \leq \mathrm{IPAR}_{\mathrm{II}} \leq 3 \\
& =2 \text { for } 4 \leq \mathrm{IPAR}_{\mathrm{II}} \leq 8 \\
& =3 \text { for } 9 \leq \mathrm{IPAR}_{\mathrm{II}} \leq 15 \\
& =4 \text { for } 16 \leq \mathrm{IPAR}_{\mathrm{II}} \leq 21
\end{aligned}
$$

The SEARCH subroutine takes the IPAR ${ }_{\text {II }}$ entry which corresponds to the subscript of the $X$ vector and seeks the corresponding column of the appropriate search matrix. This column is then searched until the currently used part number is found $\left(I P N_{J J J}\right)$. Once this occurs the subroutine increments either down or up one location depending on whether IDEX is +1 or -1 and replaces the old part number with the new one found. If the old part number happens to be on a
boundary such that a new part number cannot be obtained, the subroutine sets IBOUND to 1 and returns with the old part number. If this occurs, no further minimization can be obtained considering the $I P A R_{\text {II }}$ parameter, therefore one returns the part numbers to the base point and increments to the next most significant parameter by increasing the sub-search progress number (II) by 1 and step 6 is repeated.
7) For each new component selected by SEARCH a library subroutine, named LIBR, is called to retrieve the corresponding parameter data. This is accomplished by the statement

$$
\text { CALL LIBR[IPN } \left.{ }_{\mathrm{JJJ}}, \mathrm{XMAX}, \mathrm{XMIN}\right]
$$

where

$$
\begin{aligned}
& \mathrm{IPN}_{\mathrm{JJJ}}= \text { part number for which data is desired } \\
& \mathrm{XMAX}= \text { a vector containing the mean }+3 \text { sigma values } \\
& \text { for the total } X \text { parameter vector } \\
& \mathrm{XMIN} \mathrm{=} \mathrm{a} \mathrm{vector} \mathrm{containing} \mathrm{the} \mathrm{mean}-3 \text { sigma values } \\
& \text { for the total } X \text { parameter vector }
\end{aligned}
$$

The LIBR subroutine takes the part number ( IPN $_{J J J J}$ ) and searches the appropriate component library, stored off-line on magnetic disk, until the part number is located. Once located its associated parameter data is read back and inserted in the proper locations of the XMAX and XMIN vector. Thus by calling the LIBR subroutine with a part number, one is able to automatically update the 3 sigma limits for the X's corresponding to that part leaving the others unchanged.
8) After the new data is obtained for the exploratory move, the program checks for the existence of two conditions before the system performance is evaluated. The first is used to control the extent that the program explores changes based on a given parameter before it moves on to the next parameter. This is accomplished by calculating a normalized distance (DIST) according to

$$
\begin{align*}
& \text { DIST }=\frac{\operatorname{XMIN}_{i}}{\operatorname{XMAXS}_{i}} \text { for IDEX }{ }_{i}>0 \\
&=\frac{\operatorname{XMINS}_{i}}{\operatorname{XMAX}_{i}} \text { for IDEX }  \tag{3.49}\\
& i
\end{align*}
$$

where $i=$ IPAR $_{\text {II }}$
XMAXS $=$ a vector containing the mean +3 sigma values for the total $X$ parameter vector for the base point. XMINS $=$ a vector containing the mean -3 sigma values for the total $X$ parameter vector for the base point.

This normalized distance is then compared to a program input parameter XNN. For XNN $>1$, one is assured that the $X_{I P A R}^{I I}$ random variable has been varied so that its frequency distribution inside the 3 sigma limits lies outside the distribution for the corresponding base point parameter. Thus by selecting the value of XNN , the program user can control the extent to which exploratory moves are made. A value of $\mathrm{XNN}=1.5$ was found to give satisfactory results
and is used for the examples presented in this thesis. By making XNN larger one explores more possibilities at the expense of increased computer time. Thus for DIST > XNN the program returns the part numbers to the base point, increments to the next most significant parameter by incrementing the sub-search progress number by 1 , and returns to step 6 above by calling SEARCH. If DIST < XNN, the program continues to make the second check.

This second check consists of calculating the estimated change in the object function based on its first derivative vector using the equation

$$
\begin{equation*}
\text { sobject }=\sum_{i=1}^{m} \frac{\partial \text { object }^{\partial X_{i}}}{\left[\text { XNOM }_{i}-\text { XNOMS }_{i}\right]} \tag{3.50}
\end{equation*}
$$

where $X N O M$ and $X N O M S$ are the mean component parameter vectors corresponding respectively to the exploratory part number vector and the base point part number vector. Since the $i=I P A R_{I I}$ term in (3.50) is negative, one knows that if $\Delta$ object turns out to be positive, the summation of the changes caused by the parameters in $I^{2} N_{J J J}$ other than $I P A R_{\text {II }}$ have resulted in an estimated increase in the object function. Since an increase in $\Delta$ object is undesirable, one returns to step 6 above, when $\Delta$ object $>0$ and calls SEARCH keeping the same sub-search progress number (II). If $\Delta$ object $\leq 0$, a complete system performance analysis is made using the exploratory move part numbers.
9) If the exploratory move turns out to be "a success" (i.e., the object function is reduced) one returns to step 2 above and the process is repeated. If it is "a failure" (i.e., the object function is not reduced) one returns to step 6 and the next exploratory move is investigated.
10) The optimization procedure terminates once all the exploratory moves made from a given base point are completed "without success." This base point defines the local minimum.

Figure 3.2 is a simplified logic flow diagram for the total design program. For simplicity sake, only the logic fundamental to the directed search option is included. The path used to update the component libraries, and to calculate and store the search matrices is shown by the single dashed line. The linkage between the design program and the component and search matrix libraries via the above subroutines is illustrated with the double dashed lines.

In order to describe the operation of the program, the following additional program logic variables must be defined:

MODE $=1$ for the first analysis using the initial part numbers for each search
$=2$ for all following analyses in coarse search mode
$=3$ for all following analyses in fine search mode
$=4$ for final analysis of each search
ICOUNT $=$ number of analyses that have been conducted as part of each search

ISER = search number


Figure 3.2. Design program simplified logic diagram.

NUMRUN = specified number of searches to be made in an attempt to find the best design

IMAX = a maximum allowable iterations per search

A design begins by the user inputting the system specifications, labor cost, and the required program logic after which the program initializes numerous parameters as shown in Figure 3.2. The program then branches according to whether the input is to be from cards or selected at random. Once the part numbers are obtained, the program is at point 20 and the desired component data is retrieved from the component library using the LIBR subroutine. Since MODE was initialized $=1$, the program branches to point 200 and the system performance is calculated, the component parameters are then perturbated one at a time, and the system performance is re-evaluated. This process is repeated, using the steps described in Section 2.4, until all the entries in the Jacobian have been calculated after which the program calculates the rejection ratio, scalar, and cost (providing the later is finite).

At this point, a decision is reached whether or not to make an intermediate printout. In either event, ICOUNT is incremented by 1 and a check is made to see if it equals the maximum allowable value. Assuming the answer to be no, the program goes to the mode direction block. Since MODE $=1$, it branches to point 1 and sets MODE $=2$ (coarse search), branches to 41 and checks to see if the rejection percentage is less than 100. Assuming the answer to be no, the program stays in the coarse search mode, sets OBJECT = SCALAR and compares OBJECT to SAVE. If OBJECT < SAVE, as it is the first
time and thereafter anytime a lower object has been found, the last analysis is considered to be "a success" and the program branches to point 54 in order to store the data as a new 'base point." For the cases when OBJECT > SAVE, the last analysis is considered to be "a failure" and the program continues on with the search using the old base point via point 58.

At point 54, the new base point is established by storing the part numbers in a vector named IPS, the object function as SAVE, and the mean $\pm 3$ sigma component parameter vectors as vectors XMAXS and XMINS respectively.

Following the establishment of each new base point, the elements of the $\frac{d R}{d Y}$ vector are calculated using (3.35) or (3.37) along with the object derivative vector which in the coarse search mode equals the scalar derivative given by (3.44). The priority and direction vectors IPAR and IDEX are then established as defined by (3.47) and (3.48). After this, the sub-search progress number (II) is initialized to unity and the program is at point 57 of Figure 3.2 ready to begin a sub-search by making exploratory moves to look for a smaller object function.

At point 57, the program checks to see if $\partial$ object $/ \partial X_{I P A R}^{I I}=0$. If it is zero for a given progress number II, the program declares the base point to be a local minimum. This is accomplished by branching to point 63 , setting MODE $=4$ and returning to point 20 to terminate this search by repeating the best run obtained.

Consider now the case where for any given progress number II, the derivative is not zero. The program branches to point 58 and the SEARCH subroutine is called to select a new part number. Upon return from SEARCH the parameter IBOUND is examined to determine if one is at a boundary condition. If the answer is yes, no further minimization can be obtained considering the IPAR $_{\text {II }}$ parameter, therefore the program branches to point 59, returns the part numbers to the base point and increments to the next most significant parameter by increasing the progress number II by 1 . If II is then greater than 21 , the program has exhausted all possible parameters and therefore branches to 63 declaring the base point as the local minimum and terminates the search. If II $\leq 21$, the program returns to point 57 and the exploratory moves are continued with the next least significant parameter.

Returning to point 40 and assuming that the new part number selected by the SEARCH subroutine is not on a boundary, the program branches to point 20 and the new component data is retrieved from the library and the program branches to point 64.

At point 64, the program strategy checks for the existence of two conditions before an analysis is made. The first is to compute DIST as given by (3.49) and compare it to the program input parameter XNN. If DIST > XNN, the program branches to point 59, the part numbers are returned to the base point and the sub-search progress number is incremented to explore based on the next parameter. If DIST < XNN, the program branches to point 67 for the second check by calculating $\Delta$ object (3.50). If $\Delta o b j e c t \leq 0$, the program branches
to point 200 and an analysis is made, while if it is not, it branches to 58 and the SEARCH subroutine is called to make another part number selection.

The above described coarse search procedure is repeated until $R$ becomes less than $100 \%$ at which time the program branches to point 32 and sets $M O D E=3$ (for fine search) and OBJECT = COST and continues on as before at point 54, the only difference being that the object derivative vector is taken as the cost derivative (3.32).

The program continues in the fine search mode until either the object derivative is equal to zero or all m parameters (this case $m=21$ ) have been considered and the search is terminated via point 63. Execution of this termination is obtained by setting MODE $=4$ and repeating the system analysis using the part numbers resulting in the lowest cost for this search (i.e., the base point), the latter being defined as a local minimum.

Since this time MODE $=4$, the program branches to point 43 and ISER is compared to NUMRUN. As long as the number of searches is less than the number requested, the program branches to point 73 where the cost is compared to the best local minimum. If the new cost is less, the vector IPBEST is updated with the new part numbers and BEST is updated with the corresponding cost. In any event, ISER is incremented by 1 and a new search is started. This process is repeated until ISER $=$ NUMRUN whereupon the vector IPN is set equal to IPBEST and the analysis is repeated for the best local minimum. Upon exiting at point 43 (ISER > NUMRUN), the final printout is obtained at point 36 . The results obtained using the above described program are described in the next section.

## 4. EXAMPLE DESIGN PROBLEMS


#### Abstract

4.1 COMPONENT LIBRARIES AND SEARCH MATRICES

The components selected to make up the libraries for this study, chosen so as to provide a broad base of design, are typical of those used throughout the serwomechanism industry. The actual component parameter values used are listed in Tables 4.1 through 4.4 which consist of the "component libraries." Referring to these tables, the design problem is simply explained as "picking the one part from each table such that when combined in a system, they meet a given specification at minimum dollar cost."


Each column of the library data is labeled with the appropriate $X$-vector notation; i.e., $X_{1}, X_{2}, \ldots, X_{21}$ each of which is assumed to be a random variable with a normal distribution defined for each component by the mean $\pm 3$ sigma limits given by the $\operatorname{MAX}$ and MIN values shown. The variables $X_{i}$ for $i=1,4,9,16$, and 21 , which are the individual component costs and the gear ratio and have no manufacturing tolerance, are still treated as "random variables" but having zero variance; i.e., $X_{M A X}^{i}=X M I N_{i}$. It should be noted that many of the numerical units of measure for the variables are purposely included as an inconsistent set (e.g., min of arc, rpm, oz-in, $\mathrm{gm}_{\mathrm{cm}}{ }^{2}$, etc.). This is done to place them in one-to-one correspondence with what is normally given in vendor catalogs and component specification sheets.

Table 4.1. Followup library data.

|  | X VECTOR NOTATION |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\mathrm{X}_{1}$ | $\mathrm{X}_{2}$ |  | $\mathrm{X}_{3}$ |  |
| PART NO. | $\begin{aligned} & \text { COST } \\ & \text { DOLLARS } \end{aligned}$ | FOLLOWUP GAIN (VOLTS/RAD) MAX. $\qquad$ |  | ACCURACY (MIN OF ARC) MAX MIN |  |
| 1001 | 300.00 | 23.600C | 21.4000 | 1.0 | 0.0 |
| 1002 | 24.00 | 12.700C | 10.3000 | 10.0 | 0.0 |
| 1003 | 35.00 | 24.80CC | 2 C .2000 | 7.0 | 0.0 |
| 1004 | 200.00 | 0.5050 | 0.495 C | 30.0 | 0.0 |
| 1005 | 600.00 | 0.5025 | 0.4975 | 10.0 | 0.0 |
| 1006 | 28.00 | 24.8000 | 20.2000 | 15.0 | 0.0 |
| 1007 | 40.00 | 12.100C | 10.9000 | 3.0 | 0.0 |
| 1008 | 36.00 | 12.100 C | 10.9000 | 7.0 | 0.0 |
| 1009 | 22.00 | 12.7000 | 10.3000 | 15.0 | 0.0 |
| 1010 | 30.00 | 0.5050 | 0.4950 | 120.0 | 0.0 |
| 1011 | 95.00 | 11.7000 | 11.3000 | 2.0 | 0.0 |
| 1012 | 90.00 | 0.5050 | 0.4950 | 60.0 | 0.0 |
| 1013 | 300.00 | 0.5050 | 0.4950 | 15.0 | 0.0 |
| 1014 | 60.00 | 24.850C | 20.2000 | 3.0 | 0.0 |
| 1015 | 16.00 | 12.7c0C | 10.3000 | 30.0 | 0.0 |
| 1016 | 30.00 | 25.9000 | 19.1000 | 10.0 | 0.0 |
| 1017 | 260.00 | 11.7000 | 11.3000 | 1.0 | 0.0 |
| 1018 | 150.00 | 23.6000 | 21.4000 | 2.0 | 0.0 |
| 1019 | 20.00 | 27.0000 | 18.0000 | 30.0 | 0.0 |
| 1020 | 28.00 | 0.515 C | 0.5050 | 180.0 | 0.0 |
| 1021 | 26.00 | 5.5000 | 4.5000 | 10.0 | 0.0 |
| 1022 | 30.00 | 5.25 CO | 4.7500 | 5.0 | C. 0 |
| 1023 | 20.00 | 5.5000 | 4.5000 | 15.0 | 0.0 |
| 1024 | 28.00 | 5.2500 | 4.7500 | 7.0 | 0.0 |
| 1025 | 18.00 | 5.5000 | 4.5000 | 30.0 | 0.0 |

Table 4.2. Amplifier library data.

|  | X VECTOR NOTATION |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\mathrm{X}_{4}$ | $\mathrm{X}_{5}$ |  | $\mathrm{X}_{6}$ |  | $\mathrm{X}_{7}$ |  | $\mathrm{X}_{8}$ |  |
| PART NO. | $\begin{gathered} \cos \\ \text { ODLLARS } \end{gathered}$ | AMPLIFIF <br> TLI FOLLO <br> (VOLTS) <br> MAX. | $\begin{aligned} & \text { R GAIN } \\ & \text { WUP } \\ & \text { VGLTI } \\ & \text { MIN. } \\ & \hline \end{aligned}$ | $\begin{gathered} \text { AMPLIf I } \\ \text { TO GENE } \\ \text { IVOL TS } \\ \text { MAX. } \end{gathered}$ | $\begin{aligned} & \text { R CAIN } \\ & \text { AIIR } \\ & \text { VOLII } \\ & \text { MINO } \\ & \hline \end{aligned}$ | $\begin{aligned} & \text { SAT © } \\ & \text { I VOL T } \\ & \text { MAX. } \\ & \hline \end{aligned}$ | EVEL MIN. | $\begin{gathered} \text { OUTPU } \\ \text { IVOL } \\ \text { MAX. } \end{gathered}$ | NULL <br> $\mathrm{N}_{2}$ |
| 2001 | 20.00 | 12. | 8. | 12. | 8. | 20.0 | 16.0 | 1.00 | 0.0 |
| 2002 | 75.00 | 55. | 45. | 110. | 90. | 25.0 | 17.0 | 0.50 | 0.0 |
| 2003 | 85.00 | 1150. | 850. | 1150. | 850. | 78.0 | 20.0 | 2.00 | 0.0 |
| 2004 | 200.00 | 55. | 45. | 11000 | 9000. | 28.0 | 20.0 | 3.00 | 0.0 |
| 2005 | 140.00 | 11. | 9. | 1100. | 900. | 25.0 | 17.0 | 1.00 | 0.0 |
| 2006 | 250.00 | 105. | 95. | 10500. | 9500. | 28.0 | 20.0 | 2.00 | 0.0 |
| 2007 | 170.00 | 5250. | 4750. | 10. | 9. | 26.0 | 18.0 | 1.00 | 0.0 |
| 2008 | 107.00 | 550. | 450. | 55. | 45. | 25.0 | 17.0 | 1.00 | 0.0 |
| 2009 | 220.00 | 5500. | 4500. | 11000. | 9000. | 28.0 | 24.0 | 2.00 | 0.0 |
| 2010 | 150.00 | 11500. | 8500. | 575. | 425. | 2月.0 | 20.0 | ?. 00 | 0.0 |
| 2011 | 160.00 | 55. | 45. | 5500. | 4500. | 26.0 | 18.0 | 1.00 | 0.0 |
| 2012 | 90.00 | 105. | 95. | 53. | 48. | 25.0 | 17.0 | 1.00 | 0.0 |
| 2013 | 185.00 | 1050. | 950. | 5250. | 4750. | 28.0 | 20.0 | 1.00 | 0.0 |
| 2.014 | 180.00 | 105. | 95. | 5250. | 4750. | 26.0 | 20.0 | 2.00 | 0.0 |
| 2015 | 50.00 | 55. | 45. | 11. | 9. | 23.0 | 17.0 | 2.00 | 0.0 |
| 2016 | 75.00 | 11. | 9. | 110 | 90. | 23.0 | 17.0 | 0.50 | 0.0 |
| 2017 | 175.00 | 5500. | 4500. | 110. | 90. | 28.0 | 20.0 | 1.00 | 0.0 |
| 2018 | 220.00 | 1100 | 900. | 11000. | 9300. | 28.0 | 24.0 | 2.00 | 0.0 |
| 2019 | 180.00 | 11000. | 9000. | 11. | 9. | 26.0 | 18.0 | 1.00 | 0.0 |
| 2029 | 170.00 | 11000. | 9000. | 1100. | 900. | 28.0 | 24.0 | 2.00 | 0.0 |
| 2021 2022 | 200.00 170.00 | 550. 1050. | 450. 950. | 11000. 105. | 9000. 95. | 28.0 76.0 | 20.0 18.0 | 3.00 1.00 | 0.0 0.0 |
| 2023 | 250.00 | 10. | 9. | 10500. | 9500. | 26.0 | 18.0 | ?.00 | 0.0 |
| 2024 | 60.00 | 110. | 90. | 11. | 9. | 23.0 | 17.0 | 2.00 | 0.0 |
| 2025 | 180.00 | 525. | 475. | 525. | 475. | 26.0 | 18.0 | 1.00 | 0.0 |
| 2026 | 142.00 | 55. | 45. | 1100. | 900. | 76.0 | 18.0 | 1.00 | 0.0 |
| 2027 | 170.nc | 5500. | 4500. | 550. | 450. | 78.0 | 20.0 | 1.00 | 0.0 |
| 2028 | 130.00 | 1100. | 900. | 55. | 45. | 26.0 | 18.0 | 1.00 | 0.0 |
| 2029 | 170.00 | 550. | 450. | 5500. | 4500. | 28.0 | 20.0 | 2.00 | 0.0 |
| 2030 | 170.00 | 11000. | 9000. | 5500. | 4500. | 78.0 | 24.0 | 3.00 | 0.0 |
| 2031 | 185.00 | 11000 | 9000. | 53. | 45. | 2R. 0 | 20.0 | 1.00 | 0.0 |
| 2032 | 130.00 | 110. | 90. | 1100. | 900. | 26.0 | 18.0 | 2.00 | 0.0 |
| 2033 | 165.00 | 5500. | 4500. | 1100. | 900. | 28.0 | 20.0 | 2.00 | 0.0 |
| 2034 | 70.00 | 53. | 48. | 53. | 48. | 23.0 | 17.0 | 2.00 | 0.0 |
| 2035 | 130.00 | 53. | 48. | 525. | 5. | 25.0 | 17.0 | 1.00 | 0.0 |
| 2036 | 165.00 | 5500. | 4500. | 55. | 45. | 26.0 | 18.0 | 1.00 | 0.0 |
| 2037 | 110.00 | 550. | 450. | 1100. | 900. | 28.0 | 20.0 | 3.00 | 0.0 |
| 2038 | 105.00 | 550. | 450. | 11. | 9. | 25.0 | 17.0 | 1.00 | 0.0 |
| 2039 | 160.00 | 11. | 9. | 5500. | 4500. | 26.0 | 18.0 | 3.00 | 0.0 |
| 2040 | 110.00 | 550. | 450. | 110. | 90. | 26.9 | 18.0 | 1.00 | 0.0 |
| 2041 | 170.00 | 5500. | 4500. | 5500. | 4500. | 28.0 | 24.0 | 1.00 | 0.0 |
| 2042 | 160.00 | 11000. | 9000 . | 110. | 90. | 28.0 | 20.0 | 2.00 | 0.0 |
| 2043 | 85.00 | 11. | 8. | 575. | 425. | 25.0 | 17.0 | 2.00 | 0.0 |
| 2044 | 100.0n | 110. | 90. | 550. | 450. | 26.0 | 18.0 | 2.00 | 0.0 |
| 2045 | 160.00 | 1050. | 950. | 10. | 9. | 25.0 | 17.0 | 1.00 | 0.0 |
| 21746 | 70.00 | 110. | 90. | 110. | 90. | 25.0 | 17.0 | 1.00 | 0.0 |
| 2047 | 30.00 | 12. | $8{ }^{8 .}$ | 60. | 40. | 23.0 | 17.0 | 1.00 | 0.0 |
| 2049 | 130.00 | 1100 | 900. | 550. | 450. | 28.0 | 20.0 | 2.00 | 0.0 |
| 2049 2050 | 300.00 250.00 | 10500. 306. | 9500. | 10500. 306. | 9500. | 32.0 26.0 | 24.0 20.0 | ?. .00 $? .00$ | 0.0 0.0 |

Table 4.3. Motor-generator library data.

\begin{tabular}{|c|c|c|c|c|c|c|c|c|c|c|c|c|}
\hline \& \multicolumn{11}{|l|}{x VECtor notation} \& \\
\hline \& \(\mathrm{x}_{9}\) \& \(\mathrm{x}_{10}\) \& \(\mathrm{x}_{11}\) \& \& \& \(\mathrm{x}_{1}\) \& \& \& \(\mathrm{X}_{14}\) \& \& \(\mathrm{X}_{15}\) \& \\
\hline P4RT NO. \& \[
\begin{array}{c|c|}
\hline \operatorname{cosT} \\
\operatorname{DOLLARS}
\end{array}
\] \& \[
\begin{gathered}
\text { STALL-TCRGUE } \\
\text { COUNCEEINCHES } \\
\text { MAX. } \\
\hline
\end{gathered}
\] \& \[
\begin{array}{|l|}
\hline \text { NC-LOAC SPEEU } \\
\text { RAEV/KIN) } \\
\text { NAX. } \\
\hline
\end{array}
\] \& \({ }_{\text {ckich }}^{164}\) \& \[
\begin{aligned}
\& T 1 A \\
\& M S O R I \\
\& M I N \\
\& \hline
\end{aligned}
\] \&  \& \[
\begin{aligned}
\& \mathrm{vant} \\
\& \mathrm{TSOL} \\
\& \text { MiN. }
\end{aligned}
\] \&  \& \[
\begin{aligned}
\& 70 R \text { GAIN } \\
\& 00 \text { RPM, } \\
\& \text { MIN. }
\end{aligned}
\] \& \[
\begin{gathered}
G \in \operatorname{NER} \\
\text { MIL } \\
M A X
\end{gathered}
\] \&  \& \[
\begin{aligned}
\& \text { CONTROL } \\
\& \text { VATAGE } \\
\& \text { (RATEDI }
\end{aligned}
\] \\
\hline 3301 \& 4 \& C.13ne j.1Jce \& 157 CO .143 CC. \& \(0.1{ }^{1}\) \& \(0 \cdot 88 \mathrm{C}\) \& 0.80 \& 0.40 \& \(0 \cdot 0\) \& C.C \& \& \(0 \cdot \mathrm{C}\) \& \({ }^{38.30}\) \\
\hline 33022 \& \(32.0 n\)
39.00 \& 0.2 ccc 0.1000 \& \({ }^{\text {ragac. }} 700 \mathrm{C}\). \& 1.330 \& 0.45c \& 2.50 \& 1.25 \& - 36230 \& C. 27630 \& 1 c .00 \& 0.0 \& 26.00 \\
\hline (3003 \& 49 \&  \& 7000. 50000. \& - 1.150 \& 1.680
1.650 \& 2.50 \& 1.25 \& - 38577 \& C. 21939
0.31490 \& 10.00
15.00 \& 0.0
0.0 \& 26.00
26.00 \\
\hline 3305 \& 79.30 \& -.150c 0.12cc \& t600. 56.0. \& 0.930 \& 1.870 \& 1.00 \& 0.50 \& c. 21940 \& c. 16679 \& 5.00 \& 0.0 \& 33.90 \\
\hline 3006 \& 52. \& 0.6020 \& 8, 000 . bsoc. \& 1.700 \& 1.500 \& 0.90 \& 0.50 \& 0.54170 \& 0.39720 \& 7.00 \& 0.0 \& 22.90 \\
\hline 3307
3008
308 \& 29.30
40.00 \& - \&  \& 0.140
1.100 \& 0.080 \& 1.50
0.75 \& 0.65
0.35 \& 0.0 \& \({ }_{\text {c. }}^{\substack{0.31340}}\) \& O.0
\(15 . c 0\) \& -0.0 \& 26.90
20.00 \\
\hline \begin{tabular}{l}
3008 \\
\(3) 00\) \\
\hline
\end{tabular} \& 40.00
96.00 \&  \&  \& 1.c50 \& 0.93 C
0.44 c

0 \& 1. \& 0.35 \& O. 315 \& 0.31340
0.49540 \& \& -. 0 \& <br>
\hline 3317 \& 54.00 \& 0.75nc n.503c \& 890¢. 6302. \& 5.932 \& 2.008 \& 0.90 \& 0.45 \& 0.65940 \& C. 48530 \& 15.00 \& 0.0 \& 26.00 <br>
\hline 3311
3312
3 \& 53.00 \& 3.1600 0.130 C \& 7500. 76000. \& O.830 \& 0.684 \& 1.30 \& 0.0 \& -22030 \& C. 16070 \& 5.00 \& 0.0 \& 33.00 <br>

\hline | 3312 |
| :--- |
| 3013 | \& 7100 \& 0.250c 0.1900 \& 7900. 790000.5000. \& ¢, $\begin{aligned} & \text { 0.500 } \\ & \text { 900 }\end{aligned}$ \& 0.40 C

2.000
0.0 \& 1.20 \& 0.60 \& 0.03319 \& 0.0236 \& 4.80
16.00 \& 0.0 \& 24.00 <br>
\hline 3)14 \& 52.0 C \& -180c c. 11000 \& 10000. 8¢сС0. \& 1.10 \& 0.900 \& 1.50 \& 0.75 \& 0.57490 \& 0.47380 \& 15.00 \& 0.0 \& 26.00 <br>
\hline 3315 \& 33.28 \& 2.2500 $0.170 C$ \& P200. 6500. \& 0.170 \& 0.070 \& 1.31 \& 0.50 \& $0 \cdot 1$ \& \& \& 0.0 \& 26.00 <br>
\hline 3015
3015
3017 \& 35.00
67.00 \& 0.25cc
0.1293
$0.122 c c$
0.0742 \&  \& 1.800
0.100 \& 0.850 \& 1.30 \& C.50
0.50 \& 0.36830 \& C.27080 \& 40.00
0.0 \& 0.0
0.0 \& 23.00
23.00 <br>
\hline 3018 \& 75.20 \& 1.1294 0.6743 \& 7597. 6908. \& 0.120 \& 0.08 C \& 1.30 \& 0.50 \& . \& 0.0 \& \& 0.0 \& 23.00 <br>
\hline 3319 \& - \& 0.128c 0.07cs \& 7592. 6917. \& 0.110 \& 0.08 C \& 1.00 \& 0.50 \& 0.0 \& 0.0 \& . \& 0.0 \& 23.00 <br>
\hline 3327
3321 \& 125.00
35.00 \&  \& 11000. 90cc. \& 1.700 \& ${ }^{1.30 C}$ \& 1.10 \& 0 \& O. 50980
0.0
0.0 \& 0.07900
0.0 \& 12.00 \& 0.0 \& 32.00 <br>

\hline | 3321 |
| :--- |
| 3022 | \& 35.03

44.00 \& O.33Jc
$0.27 c c$
$0.277 c$
c.25cc \&  \& 1.000
$n .400$ \& $\xrightarrow{0.89}$ \& 1.30 \& 0.0 \& 0.n \& c.
0.0 \& 0.0 \& -0.0 \& - 20.30 <br>
\hline 3323 \& 43.0 C \& 0.160c 0.12cc \& 7500.5600. \& 0.860 \& 0.500 \& 1.00 \& 0.50 \& 0. 20030 \& 0.09800 \& 15.00 \& 0.0 \& 33.00 <br>
\hline 3324
$3>25$ \& ${ }_{4}^{36.06}$ \&  \&  \& P. 600
$0.2 n 0$ \& 1.500
0.100 \& 2.50 \& 1.25 \& \& 0.0 \& 0.0
0.0 \& 0.0 \& 28.00
15.00 <br>
\hline
\end{tabular}

Table 4.4. Geartrain library data.

|  | X VECTOR NOTATION |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $\mathrm{X}_{16}$ | $\mathrm{X}_{17}$ |  | $\mathrm{X}_{1}$ |  |  |  | $\mathrm{X}_{20}$ |  | $\mathrm{X}_{21}$ |
| PART NO. | $\begin{gathered} \text { COST } \\ \text { DOLLARS } \end{gathered}$ | $\begin{aligned} & \text { INERT } \\ & \text { IGM-CM } \end{aligned}$ | $\begin{aligned} & \text { TIA } \\ & \text { MSORI } \end{aligned}$ | STIFF IOUNCE- | NESS INCHESI | FRIC IOUNCE | CHESI | $\begin{array}{r} \text { BACKL } \\ \text { MIIN OF } \end{array}$ | $\begin{aligned} & \text { ASH } \\ & \text { ARC) } \end{aligned}$ | $\begin{aligned} & \text { GEAR } \\ & \text { RATIO } \end{aligned}$ |
|  |  | max. | MIN. | Max. | MIN. | MAX. | MIN. | MAX. | MIN. |  |
| 4001 | 250.00 | 82.5 | 67.5 | 3300. | 2700. | 0.250 | 0.062 | 3.0 | 1.5 | 50.0 |
| 4002 | 50.00 | 69.0 | 37.0 | 2052. | 1548. | 0.240 | 0.040 | 15.0 | 7.5 | 50.0 |
| 4003 | 325.00 | 1040.0 | 960.0 | 26800. | 19200. | 0.500 | 0.160 | 3.0 | 1.5 | 200.0 |
| 4504 | 300.00 | 336.0 | 264.C | 5750. | 4250. | 0.300 | 0.100 | 3.0 | 1.5 | 100.0 |
| 4005 | 60.00 | 210.0 | 190.C | 2625. | 2375. | 0.400 | 0.250 | 15.0 | 7.5 | 100.c |
| 4005 | 42.00 | 220.0 | 180.0 | 26750. | 23250. | 0.800 | 0.310 | 45.0 | 22.5 | 200.c |
| 4007 | 49.00 | 222.0 | 198.C | 12840. | 11160. | 1.000 | 0.200 | 30.0 | 15.0 | 200.0 |
| 4.008 | 135.00 | 1575.0 | 1425.0 | 129600. | 110400. | 2.000 | 0.500 | 7.0 | 3.5 | 500.0 |
| 4009 | 34.00 | 60.0 | 52.0 | 2268. | 1932. | 0.180 | 0.050 | 45.0 | 22.5 | 75.0 |
| 4017 | 53.00 | 732.0 | 668.0 | 4200. | 3800. | 0.800 | 0.200 | 30.0 | 15.0 | 500.0 |
| 4011 | 65.00 | 186.C | 174.0 | 46200. | 41800. | 0.600 | 0.150 | 15.0 | 7.5 | 200.0 |
| 4012 | 350.00 | 2184.0 | 2C16.0 | 5400. | $3600 \cdot$ | 1.000 | 0.250 | 3.0 | 1.5 | 500.0 |
| 4013 | 38.00 | 76.0 | 68.0 | 3605. | 3395. | 0.200 | 0.070 | 45.0 | 22.5 | 100.0 |
| 4014 | 30.00 | 13.0 | 11.0 | 1040. | 960. | 0.050 | 0.020 | 45.0 | 22.5 | 50.0 |
| 4015 | 120.00 | 139.C | 125.0 | 35400. | 24600. | 0.500 | 0.160 | 7.0 | 3.5 | 100.0 |
| 4015 | 70.00 | 1050.0 | 95C.C | ¢890. | 6110. | 1.500 | 0.300 | 15.0 | 7.5 | 500.0 |
| 4017 | 127.00 | 451.0 | $369 . c$ | 1248. | 1152. | 0.300 | 0.070 | 7.0 | 3.5 | 200.0 |
| 4318 | 45.00 | 87.0 | 77.0 | 16800. | 13200. | 0.300 | 0.100 | 30.0 | 15.0 | 100.c |
| 4319 | 112.00 | $113 . \mathrm{C}$ | 107. C | 84800. | 75200. | 0.370 | 0.180 | 7.0 | 3.5 | 75.0 |
| 4020 | 37.00 | $26 . C$ | 24.0 | 6540. | 5463. | 0.150 | 0.030 | 30.0 | 15.0 | 50.0 |
| 4.21 | 225.00 | 193.0 | 143.C | 1456. | 1144. | 0.150 | 0.050 | 3.0 | 1.5 | 75.0 |
| 4022 | 95.00 | 52.0 | 48. C | 10700. | 9300. | 0.200 | 0.100 | 7.0 | 3.5 | 50.0 |
| 4023 | 41.00 | 8t.C | $78 . \mathrm{C}$ | 1650. | 1350. | 0.170 | 0.050 | 30.0 | 15.0 | 75.0 |
| 4024 | 55.00 | 116.C | 108. C | 8560. | 7440 - | 0.200 | 0.100 | 15.0 | 7.5 | 75.0 |
| 4025 | 46.00 | 423.C | 377.C | 6148. | 5452. | c. 900 | 0.290 | 45.0 | 22.5 | 500.0 |

The conversion factors required to go from the units shown to the consistent set, as shown in Table 2.2, are included as part of the computer program.

Also, with respect to the motor data, the effective values of stall torque ( $\mathrm{T}_{\mathrm{s}}$ ) and no-load speed ( $\dot{\theta}_{\mathrm{m}}$ ) are functions of the voltage capability of the driving amplifier. In order to account for this effect, the values used in the program are obtained from the rated condition given in the library by the equations

$$
\begin{align*}
& \dot{\theta}_{m} \text { (effective) }=\dot{\theta}_{m} \text { (rated) } \frac{E_{s a t}}{E_{c}}  \tag{4.1}\\
& T_{s} \text { (effective) }=T_{s} \text { (rated) } \frac{E_{s a t}}{E_{c}} \tag{4.2}
\end{align*}
$$

where $E_{c}$ is the rated control voltage of the motor and $E_{\text {sat }}$ is the amplifier saturation level, both of which are included as part of the library data. Although (4.1) and (4.2) are not exact [see reference (2)] they are considered adequate for this study. In addition to the above, the motor torque gain ( $\mathrm{K}_{\mathrm{m}}$ ) and damping coefficient ( $B_{m}$ ) are terms that are used by the systems engineer and are required as part of this study; however, they are normally not provided directly by the vendor. They must be calculated from what is normally provided; no-load speed ( $\dot{\theta}_{\mathrm{m}}$ ), stall torque ( $\mathrm{T}_{\mathrm{s}}$ ), and rated control voltage ( $E_{c}$ ). The equations used are

$$
\begin{align*}
& K_{m}=\frac{T_{s}(\text { rated })}{E_{c}}  \tag{4.3}\\
& B_{m}=\frac{T_{s}(\text { rated })}{\dot{\theta}_{m}(\text { rated })} \tag{4.4}
\end{align*}
$$

The search matrices are generated immediately after the library data is stored in the computer system. These search matrices are shown as Tables 4.5 through 4.8 and consist of the component part numbers arranged in an ordered array as previously explained in section 3.5.

The computer program developed as part of this study has been tested on several design problems, each employing different specification sets. Each one has met with about equal success; however, the two that have received the most comprehensive study are presented here as examples and comprise the rest of this section.

### 4.2 FIRST DESIGN EXAMPLE

In order to demonstrate the application of the program in its most comprehensive form, a customer requirement is assumed which makes use of all eight specifications. The particular set is:

1. Static accuracy $=1.0$ degrees
2. Resolution $=0.5$ degrees
3. Velocity 1 ag for $90 \mathrm{deg} / \mathrm{sec}$ input $=5$ degrees
4. Followup rate $=90 \mathrm{deg} / \mathrm{sec}$
5. Damping ratio $=0.3$
6. Null oscillation $=$ none allowed
7. Overshoot for 10 deg step $=2.0$ degrees
8. Bandwidth for 2.0 deg peak sinusoid $=5$ hertz

The assumed labor cost is $\$ 200.00$.

First to be considered are the results obtained by using the above specification set and the program operating in the Monte Carlo mode.

Table 4.5. Followup search matrix.

| COST | K-F | THETA |
| :--- | :--- | :--- |
| 1015 | 1010 | 1017 |
| 1025 | 1005 | 1001 |
| 1023 | 1004 | 1018 |
| 1019 | 1012 | 1011 |
| 1009 | 1013 | 1014 |
| 1002 | 1020 | 1007 |
| 1021 | 1024 | 1022 |
| 1006 | 1021 | 1024 |
| 1024 | 1025 | 1003 |
| 1020 | 1022 | 1008 |
| 1022 | 1023 | 1021 |
| 1010 | 1015 | 1005 |
| 1016 | 1007 | 1002 |
| 1003 | 1011 | 1016 |
| 1008 | 1009 | 1013 |
| 1007 | 1017 | 1006 |
| 1014 | 1008 | 1023 |
| 1012 | 1002 | $10 C 9$ |
| 1011 | 1014 | 1015 |
| 1018 | 1006 | 1025 |
| 1004 | 1003 | 1019 |
| 1017 | 1018 | 1004 |
| 1013 | 1016 | 1012 |
| 1001 | 1001 | 1010 |
| 1005 | 1019 | 1020 |

Table 4.6. Amplifier search matrix.

| COST | K-AF | K-AG | SAT | NULL |
| :--- | :--- | :--- | :--- | :--- |
| 2001 | 2001 | 2001 | 2001 | 2002 |
| 2047 | 2043 | 2045 | 2047 | 2016 |
| 2015 | 2047 | 2019 | 2034 | 2027 |
| 2024 | 2023 | 2024 | 2015 | 2036 |
| 2046 | 2039 | 2015 | 2024 | 2008 |
| 2034 | 2016 | 2038 | 2016 | 2028 |
| 2016 | 2005 | 2007 | 2046 | 2022 |
| 2002 | 2015 | 2012 | 2035 | 2012 |
| 2003 | 2011 | 2047 | 2005 | 2019 |
| 2043 | 2026 | 2008 | 2012 | 2001 |
| 2012 | 2034 | 2036 | $20 C 2$ | 2047 |
| 2044 | 2004 | 2034 | 2045 | 2013 |
| 2038 | 2002 | 2028 | 2043 | 2007 |
| 2008 | 2035 | 2031 | 2008 | 2031 |
| 2040 | 2032 | 2040 | 2038 | 2025 |
| 2037 | 2044 | 2022 | 2007 | 2046 |
| 2028 | 2006 | 2017 | 2028 | 2045 |
| 2048 | 2046 | $20 C 2$ | 2026 | 2038 |
| 2035 | 2014 | 2042 | 2022 | 2005 |
| 2032 | 2012 | 2016 | 2011 | 2040 |
| 2005 | 2024 | 2046 | 2040 | 2011 |
| 2026 | 2050 | 2035 | 2023 | 2041 |
| 2010 | 2029 | 2050 | $2 C 19$ | 2035 |
| 2011 | 2025 | 2027 | 2036 | 2017 |
| 2045 | 2021 | 2025 | 2044 | 2026 |
| 2042 | 2040 | 2048 | 2025 | 2015 |
| 2039 | 2038 | 2044 | 2032 | 2029 |
| 2036 | 2037 | 2010 | 2039 | 2014 |
| 2033 | 2008 | 2043 | 2014 | 2003 |
| 2030 | 2028 | 2026 | $205 C$ | 2050 |
| 2007 | 2013 | 2003 | 2031 | 2048 |
| 2029 | 2003 | 2020 | 2029 | 2024 |
| 2027 | 2048 | 2005 | 2027 | 2006 |
| 2022 | 2045 | 2037 | 2013 | 2023 |
| 2041 | 2022 | 2033 | 2048 | 2044 |
| 2020 | 2018 | 2032 | $20 C 6$ | 2010 |
| 2017 | 2007 | 2030 | 2003 | 2020 |
| 2014 | 2027 | 2029 | 2042 | 2049 |
| 2025 | 2041 | 2014 | 2021 | 2018 |
| 2019 | 2009 | 2013 | 2010 | 2009 |
| 2031 | 2036 | 2011 | 2037 | 2042 |
| 2013 | 2017 | 2041 | 2004 | 2043 |
| 2021 | 2033 | 2039 | 2017 | 2034 |
| 2004 | 2031 | 2049 | 2033 | 2033 |
| 2018 | 2030 | 2006 | 2030 | 2032 |
| 2009 | 2049 | 2023 | 2041 | 2030 |
| 2050 | 2042 | 2021 | 2020 | 2021 |
| 2006 | 2020 | 2018 | 2018 | 2039 |
| 2023 | 2010 | 2005 | 2009 | 2037 |
| 2049 | 2019 | 2004 | 2049 | 2004 |
|  |  |  |  |  |

Table 4.7. Motor-generator search matrix.

| $C O S T$ | $T-S$ | $T-M-D$ | $J-M$ | ESTART | K-G | NULL |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  |
| 3017 | 3025 | $30 C 3$ | $3 C 15$ | 3011 | 3024 | 3022 |
| 3002 | 3003 | 3005 | $3 C 19$ | 3022 | 3022 | 3001 |
| 3015 | 3019 | 3012 | $3 C 17$ | 3021 | 3007 | 3018 |
| 3016 | 3017 | 3004 | 3001 | 3004 | 3017 | 3007 |
| 3021 | 3018 | 3008 | 3018 | 3008 | 3021 | 3015 |
| 3024 | 3004 | 3011 | $30 C 7$ | 3020 | 3001 | 3025 |
| 3003 | $30 C 1$ | 3023 | 3025 | 3001 | 3015 | 3021 |
| 3022 | 3008 | 3013 | 3022 | 3010 | 3025 | 3024 |
| 3004 | 3009 | 3010 | 3012 | 3006 | 3019 | 3019 |
| 3001 | 3005 | 3015 | $30 C 9$ | 3015 | 3018 | 3017 |
| 3025 | 3014 | $30 C 6$ | 3023 | 3023 | 3012 | 3012 |
| 3023 | 3023 | 3018 | 3003 | 3005 | 3023 | 3011 |
| 3008 | 3020 | 3019 | 3011 | 3018 | 3005 | 3005 |
| 3014 | 3007 | 3017 | 3005 | 3017 | 3011 | 3006 |
| 3006 | 3011 | 3002 | 3021 | 3016 | 3003 | 3003 |
| 3011 | 3002 | 3025 | 3008 | 3019 | 3020 | 3002 |
| 3010 | 3015 | $3 C 09$ | $3 C 14$ | 3012 | 3002 | 3020 |
| 3017 | 3016 | 3022 | $30 C 2$ | 3009 | 3016 | 3009 |
| 3012 | 3012 | 3014 | $3 C C 4$ | 3025 | 3008 | 3014 |
| 3018 | 3022 | 3007 | 3016 | 3007 | 3004 | 3023 |
| 3019 | 3021 | 3021 | 3020 | 3014 | 3006 | 3010 |
| 3005 | 3013 | 3016 | 3006 | 3013 | 3009 | 3008 |
| 3009 | 3010 | 3020 | 3024 | 3024 | 3014 | 3004 |
| 3013 | 3006 | 3001 | 3013 | 3003 | 3010 | 3013 |
| 3020 | 3024 | 3024 | 3010 | 3002 | 3013 | 3016 |
|  |  |  |  |  |  |  |

Table 4.8. Geartrain search matrix.

| $C O S T$ | $J-G$ | $K-S$ | $T-G$ | $B$ | $N$ |
| :---: | :---: | :---: | :---: | :--- | :--- |
| 4014 | 4014 | 4014 | $4 C 14$ | 4001 | 4001 |
| 4009 | 4020 | 4017 | $4 \cap 20$ | 4012 | 4022 |
| 4020 | 4022 | 4021 | $4 C 21$ | 4021 | 4020 |
| 4013 | 4002 | 4023 | 4023 | 4004 | 4014 |
| 4023 | 4009 | 4002 | $4 C 09$ | 4003 | 4002 |
| 4006 | 4013 | 4009 | $4 C 13$ | 4017 | 4023 |
| 4018 | 4001 | 4005 | 4002 | 4015 | 4009 |
| 4025 | 4023 | 4001 | 4024 | 4008 | 4019 |
| 4007 | 4018 | 4013 | $4 C 22$ | 4019 | 4024 |
| 4002 | 4019 | 4010 | 4001 | 4022 | 4021 |
| 4010 | 4024 | 4012 | $4 C 17$ | 4011 | 4015 |
| 4024 | 4015 | 4004 | 4018 | 4024 | 4013 |
| 4005 | 4021 | 4025 | $40 C 4$ | 4005 | 4005 |
| 4011 | 4011 | 4020 | 4019 | 4002 | 4018 |
| 4016 | 4006 | 4016 | $40 C 5$ | 4016 | 4004 |
| 4022 | 4005 | 4024 | $4 C 15$ | 4007 | 4007 |
| 4019 | 4007 | 4022 | 4003 | 4023 | 4003 |
| 4015 | 4004 | 4007 | 4011 | 4020 | 4011 |
| 4017 | 4025 | 4018 | 4010 | 4010 | 4006 |
| 4008 | 4017 | 4003 | 4006 | 4018 | 4017 |
| 4021 | 4010 | 4006 | 4025 | 4014 | 4025 |
| $40 J 1$ | 4003 | 4015 | $40 C 7$ | 4013 | 4012 |
| 4004 | 4016 | 4011 | 4012 | 4025 | 4010 |
| 4003 | 4008 | 4019 | 4016 | 4006 | 4015 |
| 4012 | 4012 | 4008 | 4008 | 4009 | 4008 |
|  |  |  |  |  |  |

This is accomplished by instructing the computer to make multiple analysis runs selecting the component part numbers at random. Table 4.9 illustrates a typical program output by showing the first 50 lines of intermediate printout. Each line represents an analysis run and lists the cost (3.5), scalar (3.6), total reject (3.19), the four component part numbers used, and the individual specification rejection percentages $\left[\mathrm{R}_{\mathrm{i}}\right.$ using (3.16) or (3.17) for $\mathrm{i}=1, \cdots, 8$ excluding $R_{6}$ which is given by (3.27)] arranged in order as $R_{1}, R_{2}$, $\cdots, R_{8}$. For this example, the total percentage rejection was calculated using the assumption that the $Y$ 's are independent [i.e., Equation (3.19)]. As shown, 43 out of the 50 runs illustrated have $100 \%$ rejection and therefore an infinite cost (shown as **** when cost $\geq 1 . \times 10^{6}$ dollars) while run number 20 with a cost of $\$ 663.28$ and a percent rejection of 37.88 is the best of the 50 . A total of 467 Monte Carlo runs where made (about 1.5 hours of computer time on an IBM 360 model 50). The lowest cost unit found was $\$ 374.02$, with a zero percent rejection, using part numbers 1008, 2015, 3008, and 4006. Table 4.10 is the final output sheet obtained summarizing this design.

The results obtained using the program in the direct search mode now are illustrated in detail for two searches. The first, shown in Table 4.11, is a case where the initial guess fails completely to meet 4 out of the 8 required specifications, thus resulting in an infinite cost. Twenty-seven iterations are required by the program to minimize the scalar object function to the point where the cost becomes finite and the program switches from the course to the fine search mode. It should be noted that for this run and most subsequent computer runs,

Table 4.9. Intermediate Monte Carlo printout for first design example.

| RUN MO. | cis |  | PERCENT beject | COMPONENTS SELECTED F OUP AMP MOGEN GRIM |  |  |  | -00****INOIVIOUAL |  |  | SPECIFICATIUN |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 000000000 | O41E +0 | 100.00 | 1011 | 2049 | 3005 | 4005 | 100 | 0 | O |  |  |  |  |  |
|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| 2 |  | 508 | 100.00 | 1022 | 2033 | 3025 | 4020 | 0. | 0.0 | 0.0 | 0. | 100.00 | 100.00 | 100.00 | 0 |
| 3 | ** | 1.8505 .02 | 100.0 | 1023 | $2 \mathrm{C47}$ | 3017 | 4007 | 100.00 | 100.00 | 100.00 | 0.00 | 0.0 | 0.0 | 0.0 | 100.00 |
| 4 | -0****** | $2.3011+04$ | 100.00 | 1003 | 2021 | 3002 | 4012 | 0.00 | 0.0 | 100.00 | 75.53 | 0.0 | 0.0 | 0.0 | 100.00 |
| 5 | -00****** | 3. $780 \mathrm{t}+03$ | 100.00 | 1004 | 2116 | 3010 | 4010 | 51.34 | 99.91 | 100.00 | 100.00 | 0.0 | 0.0 | 0.0 | 100.00 |
| 8 | ******** | 3. 783 t +01 | 100.00 | 1019 | 2028 | 3017 | 4014 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 100.00 | 100.00 | 0.0 |
| 7 | ******** | $3.657 t+01$ | 100.00 | 1009 | $2 \mathrm{C4O}$ | 3018 | 4001 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 0.0 | 100.00 | 0.0 |
| 8 |  | 1.047E.00 | 100.00 | 1022 | 006 | 3004 | 4016 | 99.70 | 0.00 | 100.03 | 100.00 | 0.0 | 0.0 | 0.0 | 100.00 |
| 9 | - $4 * * * * *$ | 1.409E.02 | 100.00 | 1014 | 2039 | 3024 | 4021 | 33.91 | 100.00 | 0.0 | 0.0 | 100.00 | 0.0 | 100.00 | 0.0 |
| 10 | - | $3.451 t \cdot 01$ | 100.00 | 1002 | 045 | 3015 | 4022 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 63.50 | 99.98 | 0.0 |
| 11 | -0****** | 4.944500 O | 100.00 | 1702 | 2028 | 3018 | 4010 | 0.0 | 0.0 | 0.0 | 95.13 | 100.00 | 100.00 | 0.0 | 0.0 |
| 12 | -0****** | 1.636 t -02 | 100.00 | 1022 | 044 | 3016 | 4001 | 93.82 | 92.35 | 100.05 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |
| 13 | -88****** | 5.7515001 | 100.00 | 1002 | 2001 | 3020 | 4021 | 99.89 | 100.00 | 59.41 | 0.0 | 31.34 | 0.0 | 50.46 | 100.00 |
| 14 | -******* | 8.530 E 02 | 100.00 | 1018 | 2009 | 3009 | 4002 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 0.0 | 0.00 |
| 15 | -0******* | 8.438E - 04 | 100.00 | 1006 | 2042 | 3016 | 4006 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 100.00 | 100.00 | 0.00 |
| 16 | -******** | 1.186E * 04 | 100.00 | 1917 | 2014 | 3002 | 4018 | 86. 01 | 0.00 | 100.00 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |
| 17 | - ******* | 5.040E-0才 | 100.00 | 10 Cb | 2012 | 3005 | 4009 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 99.13 | 93.34 | 0.0 |
| 10 | 37928.44 | 2.266F-01 | 98.81 | 1019 | $2 \mathrm{C48}$ | 3002 | 4016 | 0.0 | 0.0 | 0.00 | 17.36 | 0.0 | 61.08 | 0.0 | 86.48 |
| 19 | -0.0* | 1.222E 0 S | 100.00 | 1008 | 2004 | 3 COO | 4005 | 99.70 | 67.82 | 100.00 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |
| 20 | 663.28 | 5.1506 *00 | 37.88 | 1021 | 2046 | 3017 | 4007 | 0.0 | 37.88 | 0.0 | 0.00 | 0.00 | 0.0 | 0.0 | 0.0 |
| 21 | -6****** | 2.3278 -02 | 100.00 | 1023 | 2034 | 3013 | 4012 | 0.65 | 95.52 | 100.03 | 100.00 | 0.0 | 0.0 | 0.0 | 100.00 |
| 22 | **ece**** | 1.221E*01 | 100.00 | 1001 | 2031 | 3015 | 4017 | 0.0 | 0.0 | 0.0 | 0.00 | 100.00 | 100.00 | 0.0 | 0.0 |
| 23 | 113664.19 | 4.1296 .00 | 99.56 | 1021 | 2046 | 3017 | 4008 | 0.0 | 0.04 | 0.0 | 99.56 | 0.0 | 0.0 | 0.0 | 0.02 |
| 24 | -000***** | $6.974 \mathrm{~F}+01$ | 100.00 | 1019 | 2016 | 3024 | 4004 | 7.18 | 100.00 | 0.0 | 0.0 | 100.00 | 0.0 | 100.00 | 0.00 |
| 25 | -******* | 2.1746 00 | 100.00 | 1005 | 2003 | 3014 | 4019 | 92.41 | 99.94 | 100.00 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |
| 26 | -******** | - .110t 033 | 100.00 | 1003 | 2011 | 3004 | 4021 | 96.05 | 0.00 | 100.00 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |
| 27 | -******** | $1.0015 * 01$ | 100.00 | 1015 | 2040 | 3013 | 4010 | 0.0 | 0.0 | 39.38 | 100.00 | 0.00 | 3.48 | 0.0 | 100.00 |
| 28 | -0.0.0* | 9.324t.00 | 103.00 | 1022 | 2050 | 3023 | 4007 | 0.0 | 0.0 | 88.29 | 0.00 | 0.00 | 0.0 | 0.0 | 100.00 |
| 29 | -******** | 1.464E -01 | 100.00 | 1022 | 2007 | 3003 | 4022 | 0.0 | 0.0 | 0.0 | 0.00 | 100.00 | 100.00 | 100.00 | 0.00 |
| 30 | -0******* | $8.316 t \cdot 04$ | 100.00 | 1016 | 2004 | 3009 | 4022 | 98.98 | 2.17 | 100.00 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |
| 31 | -0******* | $5.791 E^{+01}$ | 100.00 | 1010 | 2021 | 3018 | 4014 | 99.93 | 100.00 | 0.00 | 0.0 | 100.00 | 0.0 | 100.00 | 0.0 |
| 32 | -0******* | 1.221E+03 | 100.00 | 1018 | 2011 | 3070 | 4024 | 91.68 | 0.00 | 99.99 | 0.0 | 0.00 | 0.0 | 0.0 | 100.00 |
| 33 | 3918.51 | 2.140f:01 | 83.74 | 1021 | 2008 | 3010 | 4001 | 0.0 | 0.0 | 0.0 | 0.0 | 0.13 | 0.0 | 83.72 | 0.00 |
| 34 | -0.***** | 3.317E*00 | 100.00 | 1025 | 2012 | 3005 | 4006 | 0.00 | 45.55 | 99.63 | 1.00 | 0.0 | 0.0 | 0.0 | 100.00 |
| 35 | ******** | 8.918E * 01 | 100.00 | 1010 | 2009 | 3001 | 4020 | 66.39 | 0.0 | 0.0 | 0.0 | 100.00 | 0.0 | 100.00 | 0.0 |
| 30 | -00****** | 1.4155.04 | 100.00 | 1012 | 2032 | 3016 | 4001 | 99.97 | 100.00 | 100.00 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |
| 37 | -0******* | 2.3075004 | 100.00 | 1017 | 2 Cl 8 | 3016 | 4016 | 51.81 | 0.0 | 100.00 | 0.00 | 0.0 | 0.0 | 0.0 | 100.00 |
| 38 | -4******* | 2.190 F 05 | 100.00 | 1001 | 2023 | 3011 | 4021 | 99.93 | 100.00 | 100.00 | 0.00 | 0.0 | 0.0 | 0.0 | 100.00 |
| 39 | 183335.81 | 1.321E*UI | 94.67 | 1018 | 2040 | 3009 | 4010 | 0.0 | 0.0 | 0.0 | 18.69 | 0.0 | 98.43 | 0.0 | 0.06 |
| 40 | -04****** | 5.194 E -00 | 180.00 | 1000 | 2040 | 3022 | 4007 | 0.0 | 0.0 | 0.0 | 0.00 | 100.00 | 100.00 | 0.0 | 0.0 |
| 41 | - ******* | 1.408t 01 | 100.00 | 1003 | 2027 | 3014 | 4012 | 0.0 | 0.0 | 0.0 | 11.97 | 0.0 | 100.00 | 0.00 | 89.73 |
| 42 | 2185.52 | $2.1700^{0} 01$ | 69.71 | 1021 | 2033 | 3008 | 4021 | 0.0 | 0.0 | 0.0 | 0.0 | 0.0 | 69.71 | 0.00 | 0.0 |
| 43 | -10060*** | 1.020 t 011 | 100.00 | 1022 | 2017 | 3007 | 4011 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 99.70 | 0.00 | 0.0 |
| 44 | -******** | 4.478601 | 100.00 | 1014 | 2029 | 3019 | 4022 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 1.12 | 100.00 | 0.0 |
| 45 | -0****** | 2.3465 *02 | 100.00 | 1015 | 2012 | 3024 | 4022 | 0.00 | 0.0 | 0.0 | 0.0 | 100.00 | 100.00 | 100.00 | 0.0 |
| 46 | - + - 0 eet* | $0.449 t .00$ | 100.00 | 1020 | 2019 | 3005 | 4003 | A5.41 | 0.0 | 0.03 | 0.07 | 100.00 | 21.11 | 0.26 | 0.00 |
| 47 | 129559.62 | 4.2461 .00 | 99.56 | 1008 | 2035 | 3017 | 4008 | 0.0 | 0.00 | 0.0 | 99.50 | 0.0 | 0.0 | 0.0 | 0.0 |
| 48 | ******** | $1.430 t-02$ | 100.0 C | 1017 | 2040 | 3024 | 4024 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 100.00 | 100.00 | 0.0 |
| 49 | ******* | 1.0031 *01 | 100.00 | 1017 | 2017 | 3018 | 4007 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 | 100.00 | 0.0 | 0.0 |
| 50 | -******** | 1.1865.04 | 100.00 | 1017 | 2014 | 3002 | 4018 | A6. 81 | 0.00 | 100.00 | 0.0 | 0.0 | 0.0 | 0.0 | 100.00 |

Table 4.10. Best design obtained using Monte Carlo for first design example.

Table 4.11. Directed search with initial guess underdesigned

the intermediate printout is eliminated for all iterations where the scalar (cost when in fine search) is not reduced. These are considered "failure iterations" as is the case for numbers 6, 10,13 , and 16 through 26 for the course search in Table 4.11.

Once the program is in the fine search mode, the cost is minimized up to run number 60 where it is reduced from $\$ 1060.21$ to $\$ 340.70$. As shown, an additional 13 iterations are required, according to the termination procedure, as explained in section 3.5, in order to establish that part numbers $1015,2015,3015$, and 4013 establish a local minimum.

Table 4.12 illustrates the results obtained from the second search. This case represents the opposite condition where the initial guess is overdesigned; i.e., all eight specifications are met to the extent that the cost is higher than required. Thus for this search, the program begins in the fine search mode and the cost is minimized. After 75 iterations, the program has reduced the per unit cost from $\$ 475.00$ to $\$ 340.70$-- a savings of $\$ 134.30$ per unit! This was accomplished at a computer run time of 23 minutes on the IBM 360 model 50 .

By comparing the $\$ 374.02$ given in Table 4.10 with the $\$ 340.70$ obtained above, it is seen that the directed search provides a cost savings of $\$ 33.32$ per unit over the Monte Carlo with a computer run time of only $25 \%$ of the latter.

A total of ten such searches were made at a cost of almost 5 hours of computer time. The resulting local minimums obtained and their frequency of occurance are summarized in Table 4.13.
Table 4.12. Directed search with initial guess overdesigned


Table 4.13. Local minimums obtained for first design example.

| Number <br> Times <br> Occurred | System <br> Cost | Rejection | Component Part Numbers |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | $\$ 634.99$ | $19.37 \%$ | 1022 | $P_{1}$ | $P_{2}$ | $P_{3}$ |
| 1 | $\$ 475.00$ | $0.0 \%$ | 1003 | 2026 | 3002 | 4015 |
| 1 | $\$ 380.61$ | $0.42 \%$ | 1025 | 2003 | 3003 | 4020 |
| 2 | $\$ 346.43$ | $2.72 \%$ | 1015 | 2015 | 3007 | 4006 |
| 5 | $\$ 340.70$ | $1.09 \%$ | 1015 | 2015 | 3015 | 4013 |

Based on the results listed in Table 4.13, the system obtained using part numbers $1015,2015,3015$, and 4013 is assumed to be the best design at a cost of $\$ 340.70$ per unit. The final computer printout sheet summarizing this combination is shown as Table 4.14.

### 4.3 SECOND DESIGN EXAMPLE

The specification set for the second example is chosen such that the computer solution time per analysis is minimized thereby enabling more example runs per dollar. This is accomplished by considering a customer requirement to consist only of the first five specifications:

1. Static accuracy $=\mathbf{0 . 3 5}$ degrees
2. Resolution $=0.3$ degrees
3. Velocity lag for $300 \mathrm{deg} / \mathrm{sec}$ input $=5$ degrees
4. Followup rate $=300 \mathrm{deg} / \mathrm{sec}$
5. Damping ratio $=0.5$

Since the last three specifications are not included, the calculation of $Y_{6}, Y_{7}$, and $Y_{8}$ as well as $R_{6}, R_{7}, R_{8}$ and $S_{6}$ can be bypassed

Table 4.14. Best design obtained using directed search for first design example.

(i.e., set equal to zero). With this alteration to the program, the computer time is reduced from approximately 18.0 to 0.3 seconds per solution -- a factor of 60.

For this specification, two sets of Monte Carlo data were obtained each comprising 4000 runs. The first 50 lines of data obtained from the first set is shown as Table 4.15. As can be seen, only 4 of the 50 have a finite cost, the best being $\$ 517.00$. It should be noted that the individual rejections for the last 3 specifications are zero since no specification exists. The total rejection for this example is calculated based on the upper bound approximation; i.e., Equation (3.13). Out of the total 8000 Monte Carlo runs made, which took about 40 minutes of computer time, the lowest cost design was found to be $\$ 375.00$ obtained using part numbers $1006,2003,3002$, and 4014 with a percentage rejection of zero. A summary of this combination is shown in Table 4.16.

The results obtained using the program in the direct search mode now are illustrated in detail for three searches. The first, shown in Table 4.17, illustrates the condition where the initial guess at first hand looks like a "reasonable design"; i.e., the rejection is only 0.77\%. However, after 74 iterations in the direct search mode, the cost has been reduced from the original design value of $\$ 555.30$ to only $\$ 374.27$-- a savings of $\$ 181.03$ per unit! The computer run time was less than one minute!

Table 4.18 illustrates the opposite condition where the initial selection of part numbers yields a system that fails completely to

Table 4.15. Intermediate Monte Carlo printout for second design example.


Table 4.16. Best design obtained using Monte Carlo for second design example.

Table 4.17. Directed search with initial guess overdesigned for second design example.


Table 4.18. Directed search with initial guess underdesigned for second design example.

|  |  |  |
| :---: | :---: | :---: |
|  |  <br> 씽ㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅇㅡ <br>  <br> 1000000000000000000000000000000 <br>  <br>  <br>  <br>  <br>  <br>  <br>  <br>  <br>  <br>  <br>  |  |
|  |  |  |
|  |  |  <br>  |
|  |  |  |
|  |  |  |
|  |  | ล̃ $\sim_{\sim}^{\sim}$ |

meet 3 out of the 5 specifications. It takes the program 55 iterations to obtain a finite cost and switch to fine search and then 147 more to reach a cost of $\$ 374.27$ which is the same local minimum.

The third search is shown in Table 4.19 where this time the initial parts result in a design which fails completely to meet 5 out of the 6 specifications. After 55 iterations, the program has reduced the scalar from $59,610,000$ to 3.396 and only one specification remains a complete failure; however, this point turns out to be a local minimum and no further reduction is obtained.

Figure 4.1 illustrates the convergence of the directed searches as compared to the results obtained from the two Monte Carlo runs by showing a plot of improved system cost versus iteration number. The two directed searches reach the minimum in 74 and 202 iterations respectively while the Monte Carlo runs employ 4000 solutions each and neither has found the minimum.

A total of 15 searches were made and the local minimums found and their frequencies are summarized in Table 4.20. Based on the results listed in Table 4.20 , the system obtained using part numbers $1009,2003,3002$, and 4014 is assumed to be the best design. The final computer printout sheet summarizing this combination is shown as Table 4.21.
Table 4.19. Directed search resulting in an unsatisfactory local minimum.



Table 4.21. Best design obtained using directed search for second design example.


Table 4.20. Local minimums obtained for second design example

| Number <br> Times <br> Occurred | System <br> Cost | Rejection | Component Part Numbers |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | $\infty$ | $100 \%$ | 1016 | 2004 | 3024 | 4025 |
| 1 | $\infty$ | $100 \%$ | 1022 | 2006 | 3015 | 4017 |
| 1 | $\$ 410.01$ | $0.25 \%$ | 1023 | 2008 | 3006 | 4014 |
| 1 | $\$ 394.63$ | $1.43 \%$ | 1009 | 2003 | 3002 | 4002 |
| 1 | $\$ 394.29$ | $0.07 \%$ | 1009 | 2012 | 3006 | 4014 |
| 9 | $\$ 374.27$ | $1.41 \%$ | 1009 | 2003 | 3002 | 4014 |
| 1 |  |  |  |  |  |  |

The validity that the above $\$ 374.27$ local minimum is also tne absolute minimum can be checked, for this example, by using the procedure explained as follows. The lowest possible cost for a system made up of any collection of components is the summation of the individual component costs and the labor cost since if there are rejects, they only increase this cost. Therefore to test if a local minimum is also the absolute minimum, one need analyze only the subset of the total combinations for which

$$
\begin{equation*}
\text { labor cost }+\sum \text { component costs }<\text { local minimum } \tag{4.5}
\end{equation*}
$$

If it turns out that analyzing each system in this subset results in a total system cost higher than the local minimum being investigated, the latter is the absolute minimum.

For the above $\$ 374.27$ local minimum there are 17,835 combinations which satisfy (4.5). This number although large is much less than the 781,250 total possible combinations and it becomes a practical value when one considers the reduced solution time. The 17,835 combinations were therefore analyzed (at a cost of 1.7 hours of computer time compared to 74.4 hours for a complete exhaustive search) and each resulted in a total system cost > $\$ 374.27$ thus proving the latter to be the absolute minimum.

It should be pointed out that the above procedure is not practical for all situations as readily apparent if one considers the $\$ 340.70$ value obtained for the first design example (see Table 4.14). This time there are only 4,249 combinations which satisfy (4.5), however, with the 18 seconds required per solution a proof would take 21 hours of computer time.

## 5. CONCLUSIONS AND RECOMMENDATIONS


#### Abstract

The objective of this thesis is to develop techniques to automatically select a collection of components that satisfy a given system specification at minimum dollar cost. This objective has been realized.


The techniques developed are sufficiently general to provide an effective design tool especially when there are significant production quantities and a large number of standard components available. The application of the developed technique is demonstrated by establishing libraries for electromechanical components and writing a computer program to automatically design instrument servomechanisms. The results of this effort are most rewarding. For example starting with an initial design that satisfied a customer specification at a cost of $\$ 475.00$ per unit, the computer program in 23 minutes produces a modified set of part numbers that meet the same specification at a cost of $\$ 340.70$ per unit. This amounts to a cost savings of $\$ 134.30$ per unit, representing a $28.2 \%$ cost reduction. In a similar manner, starting with an initial design which failed completely to meet 4 out of 8 specifications, the program brings the design from the "infinite cost condition" down to the same minimum in only 60 iterations.

The optimization techniques developed here, like most others, present no guarantee that the result obtained is an absolute minimum. A method is presented, however, that is practical in some cases for testing whether or not a local minimum is indeed the absolute minimum. This is accomplished by analyzing only a subset of the total number of possibilities which, for some cases, can be reasonably small. The test is made for one of the design examples presented which establishes that the minimum found by the program is truly the absolute minimum.

Comparisons are made between designs using purely Monte Carlo techniques and directed search. These demonstrate that the latter offers a decided advantage in faster convergence to the lowest cost design. The exact convergence time, of course, depends a great deal on the closeness of the initial guess.

In meeting the primary objective, a number of other tasks are accomplished in the area of calculating nonlinear servomechanism performance. An equation is derived for calculating the allowable backlash in a servomechanism without it displaying null oscillations. Heretofore this could only be "calculated" by iterating a direct simulation of the nonlinear state equations until the critical point was found. This could literally take several hours of computer time to obtain when considering nominal values. The thought of including the component tolerance effects was, therefore, out of the question. Wit $\mathbf{h}$ the equation derived in this study, the solution is obtained, with tolerances, in 2 seconds of computer time.

Advantages similar to the above are obtained also in the area of computing the overshoot and bandwidth for the nonlinear servomechanism.

An obvious next step for future work would be to extend the development presented here to include such design specifications as weight, volume, power consumption, and reliability. This would provide the system engineer with an even more effective total design capability.

For applications where the number of parts stored in a component library becomes extremely large, one should consider the use of a "working library" selected by the system engineer from a large "standard library." With this approach, an experienced user could eliminate, on an "a priori" basis, many components undesirable for a given application. Automated procedures would need to be developed to aid in sorting out the components with the desired features.

The search technique as developed in this thesis could be possibly further improved by extending the exploratory move strategy to include simultaneous multiple component changes. However, the added complexity of the control logic required might very well outweigh any advantages gained.

## APPENDIX A <br> DERIVATION OF BACKLASH-FRICTION SLOPE EQUATION

It is well known that backlash can cause small amplitude oscillations about null. In 1947, A. Tustin [30] presented a graphical method for analyzing the stability of systems with backlash. His work was then extended by others [31], [32], [33], and [34]. However, each of these assume zero load after the backlash (i.e., the output stops instantly each time the motor reverses). For this condition, the backlash is "represented" by a simple hysteresis nonlinearity. However, it can be shown that, for the second order system with hysteresis, the sufficient condition for asymptotic stability is only that the damping ratio ( $\zeta$ ) be greater than one half [35]. But in actual practice [2] and [36], many systems oscillate with $\zeta>0.5$, which can be attributed to the fact that the hysteresis characteristic does not represent the physical facts. References [37], [38], [39], and [40] attempt to circumvent the problem by considering a load which Continues to move after the motor reverses. Each, however, is based on the assumption that the gearing has infinite stiffness (i.e., the impact is perfectly inelastic). However, J. Liversidge [41] in 1951 demonstrated with hardware that low gear train stiffness greatly aggravated the null oscillation problem. Later efforts [42], [43], and [44] demonstrate, at least via simulation, that for an accurate model, gear stiffness must be considered along with backlash.
C. H. Thomas [45] considered this as early as 1954 but made the assumption that there was zero damping at the load. Thus, the general problem, except for direct simulation, has remained unsolved.

In the last few years, there has been considerable discussion [46], [47], [48], [49], and [50] about extending the application of describing functions to handle two separated nonlinearities in a system. The validity of such an approach, of course, increases as the amount of integration effect between them is increased. Since at least one pure integration exists in every path between the backlash and friction blocks of Figure A.1, it was decided to follow this approach.

The null oscillation problem is formulated in the following manner. Consider the 2-space of backlash and friction and some point $R$ that represents the numerical values of each for a particular system (see Figure A.2). From Reference [2] one knows that there


Figure A. 2 Backlash-friction diagram

exists a straight line (A) through the backlash-friction space that is strictly a function of the system linear parameters. If $R$ lies above this line, the system has a limit cycle while if $R$ lies below this line, the system is asymptotically stable. Our problem is to find an equation for calculating the slope (m) of this line in terms of the system linear parameters. The purpose of the remaining portion of this Appendix is to accomplish this task.

Consider again the state model diagram shown in Figure A.1. Since presently only the null oscillation problem is being considered, the saturation block may be ignored (i.e., assume a gain of unity for $N_{3}$ ). By inserting for the backlash and coulomb friction blocks their effective gains $N_{1}$ and $N_{2}$, the autonomous state model can be written as

$$
\frac{d}{d t}\left[\begin{array}{c}
\dot{\theta}_{M}  \tag{A.1}\\
\theta_{M} \\
\dot{\theta}_{L} \\
\theta_{L}
\end{array}\right]=\left[\begin{array}{cccc}
-\frac{f_{T}}{J_{M}} & \frac{N_{1} K_{S}}{J_{M}} & 0 & \frac{N_{1} K_{S}-K_{T}}{J_{M}} \\
1 & 0 & 0 & 0 \\
0 & \frac{N_{1} K_{S}}{J_{L}} & -\frac{N_{2}}{J_{L}} & -\frac{N_{1} K_{S}}{J_{L}} \\
0 & 0 & 1 & 0
\end{array}\right]\left[\begin{array}{l}
\dot{\theta}_{M} \\
{ }^{2} \\
\theta_{M} \\
{ }^{2} \\
\\
\theta^{2} \\
\dot{\theta}_{L} \\
\theta_{L}
\end{array}\right]
$$

The corresponding characteristic polynomial is then given by
$D(\lambda)=\lambda^{4}+\left[\frac{f_{T}}{J_{M}}+\frac{N_{2}}{J_{L}}\right] \lambda^{3}+\left[\frac{f_{1} N_{2}+J_{M} N_{1} K_{s}+J_{L} N_{1} K_{s}}{J_{M}{ }_{L}}\right] \lambda^{2}+\left[\frac{f_{T} N_{1} K_{s}+N_{2} N_{1} K_{s}}{J_{M} S_{1}}\right] \lambda+\left[\frac{K_{1} N_{1} K_{s}}{J_{M} J_{L}}\right]$

The describing function gains $N_{1}$ and $N_{2}$ are given as functions of their respective input zero-to-peak values, $E_{1}$ and $E_{2}$, by the equations

$$
\begin{align*}
N_{1}\left(E_{1}\right) & =1-\frac{2}{\pi}\left[\sin ^{-1}\left(\frac{B}{2 E_{1}}\right)+\frac{B}{2 E_{1}} \sqrt{1-\left(\frac{B}{2 E_{1}}\right)^{2}}\right] \text { for } E_{1} \geq \frac{B}{2} \\
& =0 \text { otherwise. }  \tag{A.3}\\
N_{2}\left(E_{2}\right) & =\frac{4 T_{L}}{\pi E_{2}} \quad \text { for all } E_{2} \tag{A.4}
\end{align*}
$$

with the corresponding range and domains

$$
\begin{array}{ll}
\frac{B}{2} \leq E_{1} \leq \infty & 0 \leq E_{2} \leq \infty  \tag{A.5}\\
0 \leq N_{1} \leq 1 & \infty \geq N_{2} \geq 0
\end{array}
$$

Since $N_{1}$ and $N_{2}$ are both real and frequency independent, one can separate out terms in (A.2) and write the characteristic equation in the form

$$
\begin{equation*}
F_{4}(\lambda)+N_{1}\left(E_{1}\right) F_{1}(\lambda)+N_{2}\left(E_{2}\right) F_{2}(\lambda)+N_{1}\left(E_{1}\right) N_{2}\left(E_{2}\right) F_{3}(\lambda)=0 \tag{A.6}
\end{equation*}
$$

where

$$
\begin{align*}
& F_{4}(\lambda)=J_{M} J_{L} \lambda^{4}+f_{T} J_{L} \lambda^{3} \\
& F_{3}(\lambda)=K_{s}{ }^{\lambda} \\
& F_{2}(\lambda)=J_{M} \lambda^{3}+f_{T} \lambda^{2} \\
& F_{1}(\lambda)=J_{T} K_{s} \lambda^{2}+f_{T} K_{s} \lambda+K_{T} K_{S} \tag{A.7}
\end{align*}
$$

Substituting $j \omega$ for $\lambda$ and defining the real and imaginary parts of each coefficient

$$
\begin{align*}
& F_{4}(j \omega)=\left(J_{M} J_{L} \omega^{4}\right)+j\left(-f_{T} J_{L} \omega^{3}\right) \equiv P_{4}+j Q_{4} \\
& F_{3}(j \omega)=j\left(K_{s} \omega\right) \equiv P_{3}+j Q_{3} \\
& F_{2}(j \omega)=\left(-f_{T} \omega^{2}\right)+j\left(-J_{M} \omega^{3}\right) \equiv P_{2}+j Q_{2} \\
& F_{1}(j \omega)=\left(-J_{T} K_{s} \omega^{2}+K_{T} K_{s}\right)+j\left(f_{T} K_{s} \omega\right) \equiv P_{1}+j Q_{1} \tag{A.8}
\end{align*}
$$

Separating the real and imaginary parts of the characteristic equation

$$
\begin{align*}
& P_{4}(\omega)+N_{1}\left(E_{1}\right) P_{1}(\omega)+N_{2}\left(E_{2}\right) P_{2}(\omega)+N_{1}\left(E_{1}\right) N_{2}\left(E_{2}\right) P_{3}(\omega)=0  \tag{A.9}\\
& Q_{4}(\omega)+N_{1}\left(E_{1}\right) Q_{1}(\omega)+N_{2}\left(E_{2}\right) Q_{2}(\omega)+N_{1}\left(E_{1}\right) N_{2}\left(E_{2}\right) Q_{3}(\omega)=0 \tag{A.10}
\end{align*}
$$

Equations (A.9) and (A.10) provide two of the necessary conditions for a limit cycle in terms of three unknowns $N_{1}, N_{2}$, and $\omega$. Solving simultaneously the two equations for $N_{1}$ and $N_{2}$ in terms of $\omega$ and substituting in for the $\mathrm{P}^{\prime} \mathrm{s}$ and $\mathrm{Q}^{\prime} \mathrm{s}$, one obtains equation (A.11) for $N_{1}$ and (A.12) for $N_{2}$, in terms of $\omega$ and the linear system parameters



Equations (A.11) and (A.12) will be referred to as the frequency relationships as they give the required describing function values as a function of frequency and the linear parameters.

The third necessary condition for a limit cycle is obtained from the fact that the derivation of (A.11) and (A.12) did not place any restrictions on the amplitude requirement that must exist between $E_{1}$ and $E_{2}$, thus $N_{1}$ and $N_{2}$. From the state model diagram (Figure A.1) one can derive a transfer function relating $e_{1}$ and $e_{2}$ as

$$
\begin{equation*}
\frac{e_{1}(\lambda)}{e_{2}(\lambda)}=\frac{-1}{\lambda}\left[\frac{J_{M^{\lambda^{2}}+f_{T}}{ }^{\lambda+K_{T}}}{J_{M^{\lambda^{2}+f_{T}}}{ }^{\lambda+N_{1}} K_{S}}\right] \tag{A.13}
\end{equation*}
$$

Letting $\lambda=j \omega$ and by taking the magnitude of both sides and $E_{1}$ $E_{2}$ be the peak values of $e_{1}(j \omega)$ and $e_{2}(j \omega)$,

$$
\begin{equation*}
\frac{E_{1}}{E_{2}}=\frac{1}{\omega} \sqrt{\frac{\left(\mathrm{~K}_{\mathrm{T}}-J_{M} \omega^{2}\right.}{\left(\mathrm{N}_{1} K_{S}-\right)^{2}+\left(\mathrm{f}_{\mathrm{T}} \omega\right)^{2}} \frac{\left.\omega^{2}\right)^{2}+\left(\mathrm{f}_{\mathrm{T}} \omega\right)^{2}}{}} \tag{A.14}
\end{equation*}
$$

Solutions are now required for $E_{1}$ and $E_{2}$ in terms of $N_{1}$ and $N_{2}$. This is difficult to acconplish for $E_{1}$. Figure A. 3 shows a plot of $N_{1}$ vs. $2 E_{1} / B$ with and without the third term of


Equation (A.3). This demonstrates that both terms are significant. However, in practice, $N_{1}$ is always close to zero so one could use a slope approximation through $(1,0)$ as

$$
\begin{equation*}
N_{1}=0.4\left[\frac{2 \mathrm{E}_{1}}{\mathrm{~B}}-1\right] \tag{A.15}
\end{equation*}
$$

Using this expression and solving for $E_{1}$

$$
\begin{equation*}
E_{1}=1.25 B\left(N_{1}+0.4\right) \tag{A.16}
\end{equation*}
$$

and solving (A.4) for $E_{2}$

$$
\begin{equation*}
E_{2}=\frac{4 T_{L}}{\pi N_{2}} \tag{A.17}
\end{equation*}
$$

Substituting (A.16) and (A.17) into (A.14) and solving for $N_{2}$

$$
\begin{equation*}
N_{2}\left(\omega, B / T_{L}\right)=\frac{3.2}{\pi\left(N_{1}+.4\right) \omega\left(B / T_{L}\right)} \frac{\left(K_{T}-J_{M} \omega^{2}\right)^{2}+\left(f_{T} \omega\right)^{2}}{\left(N_{1} K_{S}-J_{M} \omega^{2}\right)^{2}+\left(f_{T} \omega\right)^{2}} \tag{A.18}
\end{equation*}
$$

Without the slope approximation, one would be required to use the more general equation

$$
\begin{equation*}
N_{2}\left(\omega, B / T_{L}\right)=\frac{4 T_{L}}{\pi \omega E_{1}} \sqrt{\frac{\left(\mathrm{~K}_{\mathrm{T}}-\mathrm{J}_{M}{ }^{2}\right)^{2}+\left(\mathrm{f}_{\mathrm{T}} \omega\right)^{2}}{\left(\mathrm{~N}_{1} \mathrm{~K}_{\mathrm{S}}-\mathrm{J}_{M^{\omega^{2}}}\right)^{2}+\left(\mathrm{f}_{\mathrm{T}} \omega\right)}} \tag{A.19}
\end{equation*}
$$

and use an iteration of the $N_{1}\left(E_{1}\right)$ equation to evaluate $E_{1}$ given $N_{1}$ - Hereafter (A.19), or the approximate form (A.18), are referred to as the amplitude requirement, since it gives the required
value of $N_{2}$ as a function of the amplitude of the ratio of backlash to friction as well as the frequency $\omega$.

Figure (A.4) shows a plot of $N_{2}$ vs $\omega$ for a typical system using equations (A.12), and (A.18) and (A.19). The amplitude curves, from (A.18) and (A.19), coincide for each value assumed for $B / T L$ thus demonstrating that the slope approximation is valid. A necessary condition for a limit cycle to exist is that both the amplitude and frequency requirement be satisfied simultaneously. Thus, the possible limit cycle conditions are represented by the intersections shown in Figure A.4. The lower intersections represent the stable limit cycle condition of interest. The frequency of the limit cycle, corresponding to each of the lower intersections, will be denoted by $\omega_{0}$. It is seen that as the backlash-friction ratio is decreased, $\omega_{0}$ decreases, which agrees with observations obtained by simulation. This phenomenon continues until a value of backlashfriction ratio is reached where the amplitude curve becomes tangent to the frequency curve. The value of the backlash friction ratio Corresponding to this tangent curve is therefore the slope $m$ of the desired backlash-friction line. Its value can be found by solving (A.18) for $B / T_{L}$ and substituting in for $N_{2}$ and $N_{1}$ using (A.11) and (A.12). Thus

where $N_{1}(\omega)$ and $N_{2}(\omega)$ are defined as in (A.11) and (A.12). Figure A. 5 shows a plot of the bracketed quantity of (A.20) as a function of $w$. In general the problem can be solved by using a simple optimization subroutine. The one selected is SGREAT which is the subroutine version of the program described in Reference [11].

Figure A. 6 illustrates transient response curves for a typical system as obtained via a direct digital computer simulation of the nonlinear state Equation (2.1). The system parameters used are identical to those listed in Figure A. 4 except that coulomb friction $\left(T_{L}\right)$ was varied for the various curves. Looking at Figure A. 5 one sees that for low values of friction the system displays a limit cycle. As the friction is increased, the only noticeable change is a slight decrease in the limit cycle frequency ( $\omega_{0}$ ) until finally one goes beyond the critical value and the limit cycle suddenly disappears. Qualitively the results agree exactly with what one would expect from the theory that has been developed (e.g., Figure A.4). Quantitatively, however, Figure A. 6 shows that the critical value of friction is between 0.1 and 0.15 oz-in, say 0.125 . The backlash value used in the simulation was 10 minutes of arc. Thus the corresponding slope of the backlash-friction curve is calculated as

$$
m=\frac{B}{T_{L}(\text { critical })}=\frac{10}{0.125} \simeq 80 \text { minutes/oz-in }
$$

as compared to the value of 49.55 determined using (A.20). The difference is, of course, mainly attributed to the describing function approximation made in the derivation of (A.20). It can be shown, however, that neglecting the harmonics is on the safe side. That


Figure A. 5 Backlash-friction ratio vs. frequency


Figure A.6. System response curves for various friction values
is, if the harmonics had been included, the values of $N_{1}$ and $N_{2}$ would be somewhat higher. This would mean that the real damping effect would be greater and the backlash effect less than calculated in our derivation. Thus the amount of true allowable backlash is always greater than that calculated by (A.20), and the method used is always on the safe side.

## APPENDIX B

DERIVATION OF NONLINEAR OVERSHOOT EQUATION

The purpose of this appendix is to derive the necessary equations for calculating the system overshoot for large step inputs. One way this could be accomplished would be a direct numerical integration solution of the nonlinear state model given by Equation (2.1). However, this direct simulation procedure is time consuming on a digital computer. In order to minimize the solution time, the nonlinear state equations are solved explicitly using piecewise linear techniques. For simplicity sake, the system is considered to have zero backlash and infinite stiffness. This approximation is justified when considering response to large inputs since these only add small oscillations about the nominal solution.

The various solutions required are visualized best by looking at the system response on the phase-plane as shown in Figure B.l. The three response curves shown were generated on an analog computer using the simulation shown in Figure B.2. Each curve is for a different initial displacement which was selected so as to demonstrate the three different types of mode switching that can take place up to the first overshoot. Since one is concerned with the response up to the first overshoot, only the portion of the space below the abscissa is of interest. This can be divided up into three regions, in region one the drive torque is negative and saturated, and in

Figure B.1. System response on phase-plane diagram

Figure B.2. Analog computer diagram used for transient analysis.
three it is positive and saturated. Region two represents the portion of the space where the system is not saturated. The piecewiselinear state equations that define the response in each region based on the entering initial conditions are now derived. Once this is accomplished, the switching logic required for an effective solution is presented.

## B. 1 SATURATED REGION NUMBER ONE

In the saturated region number one, we have a negative driving torque and with $\dot{\theta}_{o}$ also negative, the state model is:

$$
\frac{d}{d t}\left[\begin{array}{c}
\dot{\theta}_{0}  \tag{B.1}\\
\theta_{0}
\end{array}\right]=\left[\begin{array}{cc}
-2 \zeta_{M^{\omega} N} & 0 \\
1 & 0
\end{array}\right]\left[\begin{array}{c}
\dot{\theta}_{0} \\
\theta_{0}
\end{array}\right]+\left[\begin{array}{c}
\left(\mathrm{T}_{\mathrm{L}}-\mathrm{T}_{\mathrm{SAT}}\right) / \mathrm{J}_{\mathrm{T}} \\
0
\end{array}\right]
$$

where

$$
\begin{align*}
& \zeta_{\mathrm{M}}=\frac{\mathrm{B}_{\mathrm{M}}}{2 \sqrt{\mathrm{~K}_{\mathrm{T}} \mathrm{~J}_{\mathrm{T}}}}  \tag{B.2}\\
& \omega_{\mathrm{N}}=\sqrt{\frac{\mathrm{K}_{\mathrm{T}}}{\mathrm{~J}_{\mathrm{T}}}}
\end{align*}
$$

The corresponding solution is then

$$
\begin{equation*}
\theta_{0}(t)=e^{P t} \Theta_{0}(0)+\int_{0}^{t} e^{P(t-\tau)} Q d \tau \tag{B.3}
\end{equation*}
$$

where

$$
\boldsymbol{\theta}_{0}(t)=\left[\begin{array}{c}
\dot{\theta}_{0}(t) \\
\theta_{0}(t)
\end{array}\right], P=\left[\begin{array}{cc}
-2 \zeta_{M_{N}}{ }_{N} & 0 \\
1 & 0
\end{array}\right], \quad Q=\left[\begin{array}{c}
\frac{T_{L}-T_{S A T}}{J_{T}} \\
0
\end{array}\right] \text { and } \boldsymbol{\theta}_{0}(0)=\left[\begin{array}{c}
\dot{\theta}_{0}(0) \\
\theta_{0}(0)
\end{array}\right]
$$

The eigenvalues are found from the characteristic polynominal

$$
\begin{equation*}
D(\lambda)=|\lambda U-P|=\lambda\left(\lambda+2 \zeta_{M}{ }^{\omega} N\right) \tag{E.5}
\end{equation*}
$$

thus

$$
\begin{align*}
& \lambda_{1}=0 \\
& \lambda_{2}=-2 \zeta_{M^{\omega}} N \tag{E.6}
\end{align*}
$$

The state transition matrix is given by

$$
\begin{equation*}
e^{P t}=z_{11}+z_{21} e^{-2 \zeta_{M^{\omega}} N^{t}} \tag{B.7}
\end{equation*}
$$

Substituting the above into (B.3) and simplifying

$$
\boldsymbol{\theta}_{0}(t)=z_{21}\left[\boldsymbol{\Theta}_{0}(0)+\frac{Q}{2 \zeta_{M}{ }^{\omega} N}\right] e^{-2 \zeta_{M} N^{t} t}+\left[z_{11} Q\right] t+\left[z_{11} \boldsymbol{\theta}_{0}(0)+\frac{z_{21} Q}{2 \zeta_{M}{ }^{\omega} N}\right]
$$

where the constituent idempotent matrices are calculated as

$$
\begin{align*}
& z_{11}=\frac{\lambda_{2} U-P}{\lambda_{2}-\lambda_{1}}=\frac{\left[\begin{array}{cc}
0 & 0 \\
-1 & -2 \zeta_{M^{\omega}} \mathrm{N}
\end{array}\right]}{-2 \zeta_{M^{\omega} \mathrm{N}}}  \tag{B.9}\\
& z_{21}=\frac{\lambda_{1} U-P}{\lambda_{1}-\lambda_{2}}=\frac{\left[\begin{array}{cc}
2 \zeta_{\mathrm{M}^{\omega}} \mathrm{N} & 0 \\
-1 & 0
\end{array}\right]}{2 \zeta_{\mathrm{M}^{\omega} \mathrm{N}}}
\end{align*}
$$

Substituting (B.9), and (B.4) into (B.8) and simplifying
which is the desired state equation for the saturated region number 1 .

## B. 2 UNSATURATED REGION NUMBER TWO

In the unsaturated region (region number 2) for $\dot{\theta}_{0}$ negative we have the state model

$$
\frac{d}{d t}\left[\begin{array}{c}
\dot{\theta}_{0}  \tag{B.11}\\
\theta_{0}
\end{array}\right]=\left[\begin{array}{cc}
-2 \zeta \omega_{N} & -\omega_{N} 2 \\
1 & 0
\end{array}\right] \quad\left[\begin{array}{c}
\dot{\theta}_{0} \\
\theta_{0}
\end{array}\right]+\left[\begin{array}{c}
T_{L} / J_{T} \\
0
\end{array}\right]
$$

The corresponding solution is again given by

$$
\begin{equation*}
\theta_{0}(t)=e^{P t} \theta_{0}(0)+\int_{0}^{t} e^{P(t-\tau)} Q d \tau \tag{B.12}
\end{equation*}
$$

where

$$
\boldsymbol{\theta}_{0}(t)=\left[\begin{array}{c}
\dot{\theta}_{0}(t)  \tag{B.13}\\
\theta_{0}(t)
\end{array}\right], P=\left[\begin{array}{cc}
-2 \zeta \omega_{N} & -\omega_{N}^{2} \\
1 & 0
\end{array}\right], Q=\left[\begin{array}{l}
\frac{T}{J_{T}} \\
0
\end{array}\right] \text { and } \theta_{0}(0)=\left[\begin{array}{c}
\dot{\theta}_{0}(0) \\
\theta_{0}(0)
\end{array}\right]
$$

The eigenvalues are found from the characteristic polynominal

$$
\begin{equation*}
D(\lambda)=|\lambda U-P|=\lambda^{2}+2 \zeta \omega_{N} \lambda+\omega_{N}^{2} \tag{B.14}
\end{equation*}
$$

Thus

$$
\begin{align*}
& \lambda_{1}=\omega_{N}\left(-\zeta+\sqrt{\zeta^{2}-1}\right)  \tag{B.15}\\
& \lambda_{2}=\omega_{N}\left(-\zeta-\sqrt{\zeta^{2}-1}\right)
\end{align*}
$$

The state transition matrix is given by:

$$
\begin{align*}
e^{P t} & =z_{11} e^{\lambda_{1} t}+z_{21} e^{\lambda_{2} t}, \text { for } \lambda_{1} \neq \lambda_{2} \\
& =\left[U+Z_{12} t\right] e^{-\omega_{N} t}, \text { for } \lambda_{1}=\lambda_{2}=-\omega_{N} \tag{B.16}
\end{align*}
$$

Substituting each of the above into (B.8) and simplifying

$$
\begin{gather*}
\boldsymbol{\Theta}_{0}(t)=z_{11}\left[\boldsymbol{\theta}_{0}(0)+\frac{Q}{\lambda_{1}}\right] e^{\lambda_{1} t}+z_{21}\left[\boldsymbol{\Theta}_{0}(0)+\frac{Q}{\lambda_{2}}\right] e^{\lambda_{2} t}-\left[\lambda_{2} z_{11}+\lambda_{1} z_{12}\right] \frac{Q}{\lambda_{1} \lambda_{2}} \\
\text { for } \zeta \neq 1 \tag{B.17}
\end{gather*}
$$

and

$$
\begin{array}{r}
=z_{12}\left[\Theta_{0}(0)-\frac{Q}{\omega_{N}}\right] t e^{-\omega_{N} t}+\left[\Theta_{0}(0)-\frac{Q}{\omega_{N}}-\frac{z_{12} Q}{\omega_{N}{ }^{2}}\right] e^{-\omega_{N} t}+\left[\omega_{N} U+Z_{12}\right] \frac{Q}{\omega_{N}^{2}} \\
\text { for } \zeta=1 \tag{B.18}
\end{array}
$$

where the constituent idempotent matrices are calculated as

$$
\begin{align*}
& z_{11}=\frac{\lambda_{2} U-P}{\lambda_{2}-\lambda_{1}}=\frac{\left[\begin{array}{cc}
\omega_{N}\left(\zeta-\sqrt{\zeta^{2}-1}\right) & \omega_{N}^{2} \\
-1 & \omega_{N}\left(-\zeta-\sqrt{\zeta^{2}-1}\right)
\end{array}\right]}{-2 \omega_{N} \sqrt{\zeta^{2}-1}}  \tag{B.19}\\
& z_{21}=\frac{\lambda_{1} U-P}{\lambda_{1}-\lambda_{2}}=\frac{\left[\begin{array}{cc}
\omega_{N}\left(\zeta+\sqrt{\zeta^{2}-1}\right) & \omega_{N}^{2} \\
-1 & \omega_{N}\left(-\zeta+\sqrt{\zeta^{2}-1}\right)
\end{array}\right]}{2 \omega_{N} \sqrt{\zeta^{2}-1}}
\end{align*}
$$

and the constituent nilpotent as

$$
z_{12}=\left.\left[P+\omega_{N} U\right]\right|_{\zeta=1}=\left[\begin{array}{cc}
-\omega_{N} & -_{N}^{2}  \tag{B.20}\\
1 & \omega_{N}
\end{array}\right]
$$

Substituting (B.20), and (B.15) into (B.12) and simplifying

$$
\begin{aligned}
& {\left[\begin{array}{l}
\dot{\theta}_{0}(t) \\
\theta_{0}(t)
\end{array}\right]=\left[\begin{array}{l}
\omega_{N}\left(\zeta+\sqrt{\zeta^{2}-1}\right) \dot{\theta}_{0}(0)+\omega_{N}^{2} \theta_{0}(0)-\frac{T}{J_{T}} \\
-\dot{\theta}_{0}(0)-\omega_{N}\left(\zeta-\sqrt{\zeta^{2}-1}\right) \theta_{0}(0)+\frac{T_{L}}{J_{T} \omega_{N}\left(\zeta+\sqrt{\zeta^{2}-1}\right)}
\end{array}\right] \frac{e^{-\omega_{N}\left(\zeta+\sqrt{\zeta^{2}-1}\right) t}}{2 \omega_{N} \sqrt{\zeta^{2}-1}}} \\
& +\left[\begin{array}{l}
-\omega_{N}\left(\zeta-\sqrt{\zeta^{2}-1}\right) \dot{\theta}_{0}(0)-\omega_{N}^{2} \theta_{0}(0)+\frac{T_{L}}{J_{T}} \\
\dot{\theta}_{0}(0)+\omega_{N}\left(\zeta+\sqrt{\zeta^{2}-1}\right) \theta_{0}(0)-\frac{T_{L}}{J_{T} \omega_{N}\left(\zeta-\sqrt{\zeta^{2}-1}\right)}
\end{array}\right] \frac{e^{-\omega_{N}\left(\zeta-\sqrt{\zeta^{2}-1}\right) t}}{2 \omega_{N} \sqrt{\zeta^{2}-1}}+\left[\begin{array}{l}
0 \\
\frac{T_{L}}{K_{T}}
\end{array}\right] \\
& \text { (for } \zeta \neq 1 \text { ) }
\end{aligned}
$$

and

$$
\left[\begin{array}{c}
{\left[\begin{array}{l}
\dot{\theta}_{0}(t) \\
\theta_{0}(t)
\end{array}\right]=\left[\begin{array}{c}
\frac{T_{L}}{J_{T}}-\omega_{N} \dot{\theta}_{0}(0)-\omega_{N}^{2} \theta_{0}(0) \\
\dot{\theta}_{0}(0)+\omega_{N} \theta_{0}(0)-\frac{T_{L}}{\sqrt{K_{T} J_{T}}}
\end{array}\right] t e^{-\omega_{N} t}+\left[\begin{array}{c}
\dot{\theta}_{0}(0) \\
\theta_{0}(0)-\frac{T_{L}}{K_{T}}
\end{array}\right] e^{-\omega_{N} t}+\left[\begin{array}{c}
0 \\
\frac{T}{K_{T}}
\end{array}\right]} \\
\text { for } \zeta=1 \tag{B.22}
\end{array}\right.
$$

Even though (B.21) is valid for $\zeta<1$, it does involve complex arithmetic. A better form for this case is obtained directly from (B.21) by expressing the exponentials in terms of sine and cosine and simplifying resulting in

$$
\begin{align*}
& \text { for } \varepsilon<1 \tag{B.23}
\end{align*}
$$

The necessary state equations for the unsaturated region have thus been derived. These are (B.21) for $\zeta>1$, (B.22) for $\zeta=1$, and (B.23) for $\zeta<1$.

## B. 3 SATURATED REGION NUMBER THREE

The only difference between saturated region number 1 and number 3 is that the driving torque is reversed. Therefore, the
state equation for this region is obtained by simply reversing the sign of $T_{\text {SAT }}$ in (B.10). Thus for region 3 we have

## B. 4 SWITCHING MODE LOGIC

The remaining task is to tie together the state equations for the various regions in order to arrive at the value for the system overshoot. The procedure used is summarized by the flow chart of Figure B.3, the derivation of which can be explained in the following manner.

The first step is to examine whether the servo is initially saturated by comparing $\theta_{0}(0)$ to $T_{S A T} / K_{T}$. Assuming that the initial displacement is large enough to cause saturation, the next step is to solve for the intersection of the trajectory and the first saturation boundary. The boundary line defining this saturation is given by

$$
\begin{equation*}
\theta_{\mathrm{b}_{1}}=\frac{\mathrm{T}_{\mathrm{SAT}}}{\mathrm{~K}_{\mathrm{T}}}-\frac{2 \zeta_{\mathrm{g}}}{\omega_{\mathrm{N}}} \dot{\theta}_{\mathrm{O}} \tag{B.25}
\end{equation*}
$$

letting the $\left(T_{S A T}-T_{L}\right) /{ }^{B_{M}}$ term of Equation (B.10) be defined as $\dot{\theta}_{S}$, which is the servo followup rate, and then substituting (B.10) into (B.25):


Figure B.3. Nonlinear overshoot logic flow diagram.

$$
\begin{align*}
& {\left[\frac{-\dot{\theta}_{0}(0)-\dot{\theta}_{s}}{2 \zeta_{M^{\omega} N}}\right] e^{-2 \zeta_{M} \omega_{N} t}-\dot{\theta}_{s} t+\theta_{0}(0)+\frac{\dot{\theta}_{0}(0)+\dot{\theta}_{s}}{2 \zeta_{M} M_{N}}} \\
& \quad=\frac{T_{S A T}}{K_{T}}-\frac{2 \zeta_{g}}{\omega_{N}}\left[\dot{\theta}_{0}(0)+\dot{\theta}_{s}\right] e^{-2 \zeta_{M} M_{N} t}-\frac{2 \zeta_{g}}{\omega_{N}} \dot{\theta}_{s} \tag{B.26}
\end{align*}
$$

and collecting terms

$$
\begin{align*}
& {\left[\frac{2 \zeta_{g}}{\omega_{N}}-\frac{1}{2 \zeta_{M^{\omega}}{ }^{\omega}}\right]\left[\dot{\theta}_{0}(0)+\dot{\theta}_{s}\right] e^{-2 \zeta_{M}{ }^{\omega} N^{t} 1}-\left[\dot{\theta}_{s}\right] t_{1}} \\
& \quad+\left\{\frac{1}{2 \zeta_{M^{\omega}} N}\left[\dot{\theta}_{0}(0)+\dot{\theta}_{s}\right]+\theta_{0}(0)-\frac{T_{S A T}}{K_{T}}-\frac{2 \zeta_{g}}{\omega_{N}} \dot{\theta}_{s}\right\}=0 \tag{B.27}
\end{align*}
$$

Equation (B.27) can then be solved for the time at the boundary $\left(t_{1}\right)$ by iteration. A good initial guess is

$$
\begin{equation*}
t_{1} \text { (est.) }=\frac{\theta_{0}(0)-T_{S A T} / K_{T}}{\dot{\theta}_{S}} \tag{B.28}
\end{equation*}
$$

Once $t_{1}$ is found it is substituted into (B.10) to obtain the desired boundary conditions which are the initial conditions for region number 2. These two steps are illustrated as the first operation in Figure B. 3 for the saturated case.

Region number 2 is then entered, either after the above calculations for the saturated case, or directly for the unsaturated case.

Since region number 2 is one of indecision, that is the next boundary is not known, one simply continues to solve for $\dot{\theta}_{0}(t)$ and $\theta_{0}(t)$ incrementing time in steps of $\Delta t$ until one of the two boundaries is crossed. However, the particular state equation to be evaluated in region number 2 depends on the value of $\zeta$. Thus the next step is to evaluate $\zeta$ and, depending on its value, calculate the matrix coefficients which are independent of time as required for either (B.21), (B.22), or (B.23). One is now at point 5 of Figure B. 3 with $t=0$. The previous values of the state variables are stored as $\Theta_{0 S}$ and depending on the value of 5 either (B.21), (B.22), or (B.23) is evaluated. It should be noted that the latter is a relatively simple calculation since the matrix coefficients are independent of $t$ and have already been evaluated.

After point 9, one simply checks to see if either of the two possible boundary conditions have been exceeded, i.e. if $\dot{\theta}_{0}(t)>0$ or if $\theta_{0}(t)<\theta_{b 2}$ where $\theta_{b 2}$ in the corresponding values for the second saturation line given by

$$
\begin{equation*}
\theta_{\mathrm{b} 2}=-\frac{\mathrm{T}_{\mathrm{SAT}}}{\mathrm{~K}_{\mathrm{T}}}-\frac{2 \zeta_{\mathrm{g}}}{\omega_{\mathrm{N}}} \dot{\theta}_{\mathrm{O}} \tag{B.29}
\end{equation*}
$$

If neither of the above inequalities is satisfied then an appropriate $\Delta t$ is calculated, $t$ is incremented and one returns to point 5 and the process is repeated until an exit is obtained at either point 10 or 11.

Consider first that the exit is via 11. The corresponding situation is illustrated in Figure B. 4 , i.e. $\theta_{0}(t)$ lies in region


Figure B.4. Phase-plane interpolation diagram.
number 3 while $\theta_{0}(t-\Delta t)$ lies in region 2 . By using a straight interpolation line between these two points, the desired intersection point is readily calculated using the derived interpolation equation

$$
\left[\begin{array}{c}
\dot{\theta}_{0}\left(t_{2}\right)  \tag{B.30}\\
\theta_{0}\left(t_{2}\right)
\end{array}\right]=\frac{1}{2 \zeta_{g}{ }^{M+\omega_{N}}}\left[\begin{array}{l}
\omega_{N} \dot{\theta}_{0}(t-\Delta t)-\omega_{N} \theta_{0}(t-\Delta t)-\omega_{N}{ }^{M T}{ }_{S A T} / K_{T} \\
2 \zeta_{g}{ }^{M} \theta_{0}(t-\Delta t)-2 \zeta_{g} \dot{\theta}_{0}(t-\Delta t)-\omega_{N} T_{S A T} / K_{T}
\end{array}\right]
$$

where $M$ is the slope of the interpolation line given by

$$
\begin{equation*}
M=\frac{\dot{\theta}_{0}(t)-\dot{\theta}_{0}(t-\Delta t)}{\theta_{0}(t)-\theta_{0}(t-\Delta t)} \tag{B.31}
\end{equation*}
$$

In a similar manner, if one exceeds the $\dot{\theta}_{0}=0$ boundary (i.e. exit via point 10) the intersection point, which this time corresponds to the desired overshoot, may be evaluated using

$$
\begin{equation*}
\theta_{0}(\text { overshoot })=\frac{\dot{\theta}_{0}(t-\Delta t)}{M}-\theta_{0}(t-\Delta t) \tag{B.32}
\end{equation*}
$$

where $M$ is as given in (B.31).

If the intersection is on the $\dot{\theta}_{0}=0$ line, one is finished; if not, then one proceeds into region number 3 using the new initial conditions as given by (B.30). Region number 3 is also one of indecision reguarding the next boundary condition. However one can readily derive which boundary applies in the following manner. If one remained in region number $3 \dot{\theta}_{0}(t)$ would equal zero at some time say $t_{3}$. This value can be solved forby using the first equation of (B.24) yielding

$$
\begin{equation*}
t_{3}=\frac{1}{2 \zeta_{M^{\omega} N}} \ln \left[1-\frac{\mathrm{B}_{M_{0}} \dot{\theta}_{o}(0)}{\mathrm{T}_{S A T}{ }^{+T} \mathrm{~L}}\right] \tag{B.33}
\end{equation*}
$$

and using this value in the second equation one readily obtains $\theta_{0}\left(t_{3}\right)$. If this value is less than $-T_{\text {SAT }} / K_{T}$, the assumption of remaining in region number 3 is valid and one ends up at point 12 of Figure B. 3 with $\theta_{0}$ (overshoot) $=-\theta_{0}\left(t_{3}\right)$. If not then the trajectory must enter region number 2 for the second time. An equation giving the appropriate time in region 3 can be obtained in the same manner as (B.27) by letting $\dot{\theta}_{S S}=\left(T_{S A T}+T_{L}\right) / B_{M}$ and using (B.24) instead of (B.10).

$$
\begin{align*}
& {\left[\frac{2 \zeta_{g}}{\omega_{N}}-\frac{1}{2 \zeta_{M} \omega_{N}}\right]\left[\dot{\theta}_{0}(0)+\dot{\theta}_{s s}\right] e^{-2 \zeta_{M} \omega_{N} t_{3}}+\dot{\theta}_{s s} t_{3}} \\
& +\left\{\frac{1}{2 \zeta_{M} \omega_{N}}\left[\dot{\theta}_{0}(0)-\dot{\theta}_{s s}\right]+\theta_{0}(0)+\frac{T}{T_{S A T}}+\frac{2 \zeta_{T}}{\omega_{N}} \dot{\theta}_{s s}\right\}=0 \tag{B.34}
\end{align*}
$$

Once the time in region number $3\left(\mathrm{t}_{3}\right)$ is obtained by sloving (B.34) it is substituted into (B.24) and the initial conditions are obtained for region number 2, $t$ is set to $\Delta t$ and one returns to point 5 . The procedure is thus repeated as explained before except this time a termination is reached via point 10 and Equation (B.32).

Tables B.1, B.2, and B. 3 show sample outputs of a computer program that was written to perform the above described procedure. The three tabulations correspond to the three trajectories shown in Figure B.1. These trajectories are repeated as Figure B. 5 with the corresponding calculated data points from the digital computer program supperimposed. Figure B. 6 illustrates the same results except on the displacement vs. time plot instead of the phase-plane.

Table B.1. Nonlinear overshoot calculations for 0.1 radian displacement.


Table B.2. Nonlinear overshoot calculations
for 0.2 radian displacement.


Table B.3. Nonlinear overshoot calculations for 0.35 radian displacement.

$K_{T}=1536$ OZ-IN/RAD



Figure B.6. Typical system response curves.

## APPENDIX C <br> DERIVATION OF NONLINEAR BANDWIDTH EQUATION

The purpose of this appendix is to derive the necessary equations for calculating the system bandwidth. Often, this is "accomplished" by using the linear closed loop transfer function, with s replaced by $j \omega$, and calculating the frequency at which $\left|\theta_{0}(j \omega) / \theta_{i}(j \omega)\right|$ is down 3 db . However in the real world, one rarely realizes this value. This discrepancy is mainly a result of amplifier saturation, which in turn causes the actual bandwidth to be a function of input level. The true affects of saturation, as well as coulomb friction, backlash and finite stiffness, could be accounted for by direct simulation of the nonlinear state equations presented in Section 2.2. However, including this simulation in an iteration loop (as necessary to find $\omega=\omega_{B}$ such that the response is -3 db ) is very time consuming. This is due to the fact that for each iteration one must wait for steady-state conditions before an evaluation can be made. In order to minimize the solution time, an algebraic equation for the bandwidth frequency is derived including the effects of saturation and coulomb friction. This is accomplished by using describing function approximations. For simplicity sake, the system is considered to have zero backlash and infinite stiffness. This approximation is justified since the displacements are considered to be large compared to any deflections that may exist in the gear train.

The bandwidth frequency for the linear second order system is found by taking the closed loop transfer function

$$
\begin{equation*}
\frac{\theta_{0}(s)}{\theta_{i}(s)}=\frac{\omega_{N}^{2}}{s^{2}+2 \zeta \omega_{N} s+\omega_{N}^{2}} \tag{C.1}
\end{equation*}
$$

replacing $s$ by $j \omega$ and setting the magnitude equal to 0.707 for $\omega=\omega_{B}$

$$
\begin{equation*}
\left|\frac{\omega_{N}^{2}}{\left(\omega_{N}^{2}-\omega_{B}^{2}\right)+j\left(2 \zeta \omega_{N} \omega_{B}\right)}\right|=0.707 \tag{C.2}
\end{equation*}
$$

and solving for the bandwidth frequency ( $\omega_{B}$ )

$$
\begin{equation*}
\omega_{B}=\omega_{N} \sqrt{1-2 \zeta^{2}+\sqrt{2-4 \zeta^{2}+4 \zeta^{4}}} \tag{C.3}
\end{equation*}
$$

This equation can be extended to the desired nonlinear case using describing functions, by replacing $\omega_{N}$ and $\zeta$ by their corresponding effective values $\omega_{N}$ ' and $\zeta^{\prime}$ given by

$$
\begin{equation*}
\omega_{N} \prime=\sqrt{\frac{K_{f}^{K_{a}}{ }^{\prime} K_{m}}{J_{T}}} \tag{C.4}
\end{equation*}
$$

$$
\begin{equation*}
\zeta^{\prime}=\frac{\mathrm{B}_{M+K_{G} K_{a}^{\prime} / K_{a}+N_{2}}^{2 \sqrt{K_{f} K_{a}^{\prime} K_{M}{ }^{\prime} T}}}{2} \tag{C.5}
\end{equation*}
$$

where $K_{a}^{\prime}$ and $N_{2}$ are the effective gain values for the amplifier and coulomb friction elements as given by their respective describing functions. For the friction element

$$
\begin{equation*}
N_{2}=\frac{4 T_{L}}{\pi \dot{\theta}_{0}(\text { peak })} \tag{c.6}
\end{equation*}
$$

and letting $\mathrm{N}_{3}$ be the describing function for the amplifier saturation

$$
\begin{align*}
N_{3} & =\frac{2 K_{a}}{\pi}\left[\psi+\frac{\sin (2 \psi)}{2}\right] \text { for } E_{i}(\text { peak }) \geq E_{s a t} K_{a}  \tag{C.7}\\
& =K_{a} \text { otherwise }
\end{align*}
$$

where

$$
\begin{equation*}
\psi=\operatorname{arc} \sin \left[\frac{E_{\text {sat }}}{K_{a} E_{i} \text { (peak) }}\right] \tag{C.8}
\end{equation*}
$$

Thus $N_{3}$ is a function of its respective input $E_{i}$ (peak) and likewise $N_{2}$ is a function of $\dot{\theta}_{0}$ (peak). One can obtain $E_{i}$ (peak) for any $\omega$ from the transfer function

$$
\begin{equation*}
\frac{E_{i}(s)}{\theta_{i}(s)}=K_{f}\left[\frac{s^{2}+2 \zeta_{M L} '_{N} \omega^{\prime} s}{s^{2}+2 \zeta^{\prime} \omega_{N}{ }^{\prime} s+\omega_{N}{ }^{\prime 2}}\right] \tag{C.9}
\end{equation*}
$$

where

$$
\begin{equation*}
\zeta_{M L}^{\prime}=\frac{\mathrm{B}_{M+N_{2}}}{2 \sqrt{K_{f}{ }^{K_{a}^{\prime}{ }^{\prime} K_{M} J_{T}}}} \tag{C.10}
\end{equation*}
$$

by replacing $s$ with $j \omega$ and solving for $E_{i}$ (peak)

$$
\begin{equation*}
E_{i} \text { (peak) }=K_{f} \sqrt{\frac{\omega^{4}+\left(2 \zeta_{M L} \omega_{N} \omega^{\prime} \omega\right)^{2}}{\left(\omega_{N}^{\prime 2}-\omega^{2}\right)^{2}+\left(2 \zeta^{\prime} \omega_{N}{ }^{\prime} \omega\right)^{2}}} \theta_{i} \text { (peak) } \tag{C.11}
\end{equation*}
$$

Since the output amplitude is 0.707 of the input amplitude

$$
\begin{equation*}
\dot{\theta}_{0}(\text { peak })=0.707 \omega \theta_{i}(\text { peak }) \tag{C.12}
\end{equation*}
$$

and thus

$$
\begin{equation*}
N_{2}=\frac{4 \mathrm{~T}_{\mathrm{L}}}{0.707 \pi \omega \theta_{\mathrm{i}} \text { (peak) }} \tag{C.13}
\end{equation*}
$$

The necessary relationships have now been derived and $\omega_{B}$ given by

$$
\begin{equation*}
\omega_{B}=\omega_{N} \prime \sqrt{1-2 \zeta^{\prime 2}+\sqrt{2-4 \zeta^{\prime 2}+4 \zeta^{\prime 4}}} \tag{C.14}
\end{equation*}
$$

is calculated using the iteration scheme illustrated in Figure C.l. This procedure consists of one iteration inside another. The outer loop is used to adjust $\omega$ until it equals $\omega_{B}$ as given by (C.14) to the desired accuracy $\left(\delta \omega_{B}\right)$, while the inner loop adjusts, for each value of $\omega, K_{a}$ ' so that it equals its describing function value $\left(N_{3}\right)$ as given by (C.7) to the desired accuracy ( $\delta K_{a}$ ).

Tables C. 1 and C. 2 show sample outputs from a computer program that was written to perform the above described procedure. The only difference between the two tables is that Table C. 2 is for the zero friction case. Each table consists of a tabulation of the system


Figure C.1. Nonlinear bandwidth flow diagram.

Table C.1. Nonlinear bandwidth calculations with friction and amplifier saturation.


Table C.2. Nonlinear bandwidth calculations for zero friction case.

bandwidth as a function of peak magnitude of the input sine wave. There is little difference between the two tabulations thereby demonstrating that amplifier saturation is the dominate nonlinearity except for small input levels. At small input levels (less than about 4 degrees) the zero friction case is equivalent to the linear case (no saturation) and the bandwidth equals the linear value of $61.74 \mathrm{rad} / \mathrm{sec}$. For the case with friction the bandwidth falls off for both high and low values of input amplitude, as would be expected. The results obtained are plotted as Figure C.2.

An analog computer was used to check the validity of the describing function approximations used to represent the two nonlinearities. Figure C. 3 documents the simulation used. The analog computer was operated by setting potentiometers 5 and 6 to the corresponding values of $\omega_{B}$ and $\theta_{i}$ (peak) as tabulated by the digital program and making Lissajous diagrams in predrawn boxes with the height (output) equal to 0.707 times the base (input). As demonstrated by Figure C. 4 the analog computer responses are almost exactly tangent to all sides, thereby demonstrating that the describing function assumption provides an effective model of the system.


Figure C.3. Analog computer diagram used for bandwidth analysis.


Figure C.4. Lissajous diagrams obtained from analog simulation.

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