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FEASIBILITY INVESTIGATION FOR DETERMINING ARMY HELICOPTER GAS TURBINE ENGINE MAXI-MUM POWER AVAILABLE

Joseph M. Kos, et al

Hamilton Standard

#### Prepared for:

Army Air Mobility Research and Development Laboratory

February 1973

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## **USAAMRDL TECHNICAL REPORT 72-58**

# FEASIBILITY INVESTIGATION FOR DETERMINING ARMY HELICOPTER GAS TURBINE ENGINE MAXIMUM POWER AVAILABLE

By
Joseph M. Kos
Anthony J. Martin
Peter D. Miller
John Saunders
Roy W. Schneider
Louis A. Urban
Frank C. Zuliani



## **EUSTIS DIRECTORATE**

February 1973

# U. S. ARMY AIR MOBILITY RESEARCH AND DEVELOPMENT LABORATORY FORT EUSTIS, VIRGINIA

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HAMILTON STANDARD

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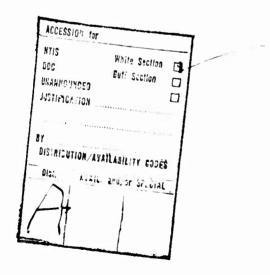
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# DEPARTMENT OF THE ARMY U. S. ARMY AIR MOBILITY RESEARCH & DEVELOPMENT LABORATORY EUSTIS DIRECTORATE FORT EUSTIS, VIRGINIA 23604

This report was prepared by the Hamilton Standard Division of United Aircraft Corporation under Contract DAAJ02-72-C-0003. The basic objective of the effort was to determine the feasibility of developing a method to predict the maximum power available (MPA) from a helicopter gas turbine engine at full-power conditions. The MPA prediction was to be made with an accuracy goal of at least  $\pm 1\%$  using information obtained from the engine while the engine was operated at a partial-power condition of no more than 30% of normal rated power.

The report consists of a discussion of the various algorithms that could be considered in determining maximum power available, and the algorithm that provides the most accurate method for predicting maximum power available is selected.

The results of the investigation show that development of a method to predict the MPA of an Army helicopter gas turbine engine prior to lift-off is feasible; however, before an MPA prediction system with an accuracy of  $\pm 1\%$  can be developed, more accurate sensors must be developed. The accuracy of such a system can also be enhanced by acquiring the necessary parametric information while operating the engine at partial-power conditions higher than 30% of normal rated power.

The conclusion and recommendations are generally concurred in by this Directorate; however, before an MPA prediction system is fabricated, it is felt that an intermediate effort should be conducted to determine the accuracy of and optimize a system that uses parametric information obtained from an engine operated at a power condition higher than 30% of normal rated power.

The technical monitor for this contract was Mr. G. William Hogg, Military Operations Technology Division.

Unclassified Security Classification DOCUMENT CONTROL DATA - R & D Security classification of title, body of abstract and indexing annotation must be entered when the overall report is classified, ORIGINATING ACTIVITY (Corpirate author) 28. REPORT SECURITY CLASSIFICATION Hamilton Standard Unclassified Division of United Aircraft Corporation Windsor Locks, Connecticut FEASIBILITY INVESTIGATION FOR DETERMINING ARMY HELICOPTER GAS TURBINE ENGINE MAXIMUM POWER AVAILABLE 4 DESCRIPTIVE NOTES (Type of report and inclusive dates) Final AUTHOR(\$) (First name, middle initial, last name) Joseph M. Kos John Saunders Frank C. Zuliani Anthony J. Martin Roy W. Schneider Peter D. Miller Louis A. Urban REPORT DATE TOTAL NO. OF PAGES 75. NO. OF REFS February 1973 118 Se. ORIGINATOR'S REFORT NUMBER(S) SE. CONTRACT OR GRANT NO DAAJ02-72-C-0003 b. PROJE : T NO USAAMRDL Technical Report 72-58 Task 1F162203A43405 9b. OTHER REPORT NO(5) (Any other numbers that may be assigned this report) O DISTRIBUTION STATEMENT Approved for public release; distribution unlimited. 12. SPONSORING MILITARY ACTIVITY 11 SUPPLEMENTARY NOTES Eustis Directorate U.S. Army Air Mobility R&D Laboratory Fort Eustis, Virginia 3 ABSTRACT

The purpose of this investigation was to determine the feasibility of developing a method to predict, with an accuracy of better than ±1%, the maximum power which can be produced by a helicopter gas turbine engine at full-power conditions. The prediction was to be made using information obtained from the engine while the engine was operated prior to lift-off at a partial-power condition of no more than 30% of normal rated power. The prediction method was to be capable of identifying the changes in maximum engine power available due to all possible types of engine deterioration and all ambient conditions. The study was based on a Lycoming T53-L13 gas turbine engine currently being used in the Army UH-1 helicopter. The feasibility study was first based on the use of information available from existing sensors normally installed on the engine. It was determined, however, that the use of these sensors did not permit the desired ±1% accuracy to be obtained, and for this reason the feasibility study was expanded to include the use of additional and/or more accurate sensors. In addition, it was assumed that input information was available which provided atmospheric density data with an accuracy of ±.75%.

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#### SUMMARY

An investigation was conducted by Hamilton Standard Division of UAC, Windsor Locks, Connecticut, to determine the feasibility of developing a method to predict the maximum power available (MPA) from a helicopter gas turbine engine at full-power conditions. The MPA prediction was to be made with an accuracy goal of at least  $\pm 1\%$  using information obtained from the engine while the engine was operated at a partial-power condition of no more than 30% normal rated power. The MPA prediction was to take into account the effects of all ambient conditions and all internal modes of engine deterioration.

The Lycoming T53-L13 engine, a gas turbine engine presently in use on the Army UH-1 helicopter, was selected for the investigation. A mathematical model of the T53-L13 engine and an MPA prediction system was developed based on Hamilton Standard's prior experience and knowledge of engine control and diagnostic systems. This model was analyzed to determine the best attainable MPA prediction accuracy assuming perfect sensors, inaccuracies, and effects on predicted MPA due to all input parameters, the effect of power condition on MPA prediction, and possible alternate MPA prediction methods using various sets of parametric sensors and making various assumptions regarding the relative values of independent engine parameters. The model was further evaluated by the use of actual engine operational test data taken by Hamilton Standard as a part of the U. S. Army UH-1 AIDAPS feasibility study. Hardware implementation of an MPA prediction system was also investigated.

Development of an MPA prediction system is feasible. An initial system accuracy of  $\pm 3.5\%$  can be achieved using an aircraft-mounted digital computer and the most accurate, aircraft type parametric sensors available today.

Attainment of the  $\pm 1\%$  accuracy goal is not feasible under the ground rules for this study. Further studies are required to determine if  $\pm 1\%$  accuracy is attainable when the MPA prediction is made at power levels higher than 30%, on multi-engine helicopters for example, and when the baseline engine data, stored in the computer, is updated during flight to reduce errors due to ambient variations and engine degradation.

#### TABLE OF CONTENTS

	Page
SUMMARY	
LIST OF ILLUSTRATIONS	vi
LIST OF TABLES	viii
LIST OF SYMBOLS	хi
INTRODUCTION	1
DISCUSSION	3
The Mathematical System Model Confirmation of Model Accuracy Parametric Sensitivity Studies and Error Analysis Effect of Power Condition on MPA Prediction Alternate MPA System Studies Final System Selection System Model Evaluation Using AIDAPS Flight Test Data System Hardware Implementation	3 20 25 58 69 75 86 91
CONCLUSIONS	106
RECOMMENDATIONS	
APPENDIXES	
I. Notes Associated With Error Summary Sheets	108
II. System Growth	112
DISTRIBUTION	115

#### LIST OF ILLUSTRATIONS

Figure		Page
1	Schematic Diagram of a Single-Spool Free-Power Turbine Engine	5
2	Compressor Efficiency Versus $N_1 \sqrt{\theta_1}$	7
3	Gas Generator Turbine Efficiency Versus Turbine Pressure Ratio	7
4	Power Turbine Efficiency Versus Turbine Pressure Ratio	8
5	Corrected Gas Generator Speed Versus Corrected Compressor Discharge Pressure	8
6	Corrected Compressor Discharge Temperature Versus Corrected Compressor Discharge Pressure	9
7	Corrected Fuel Flow Versus Corrected Compressor Discharge Pressure	9
8	Corrected Gas Generator Turbine Discharge Pressure Versus Corrected Compressor Discharge Pressure	10
9	Corrected Power Turbine Horsepower Versus Corrected Compressor Discharge Pressure	10
10	Corrected Power Turbine Discharge Temperature Versus Corrected Compressor Discharge Pressure	11
11	$T_7$ Limit Mode Ambient Temperature Correction Factor Versus $T_{AM}$	16
12	N <sub>1</sub> Limiting Mode Ambient Temperature Correction Factor Versus T <sub>AM</sub>	17
13	W <sub>f</sub> Limit Mode Ambient Temperature Correction Factor Versus TAM	18
14	$W_f$ Versus $N_1$ for Base-Line Engine and for Base-Line Engine Degraded by $1\%$ Increase in $\eta_c$ , $\eta_t$ , $\eta_{pt}$ , $\Lambda_5$ and $\Lambda_N$	21

Figure		Page
15	hppt Versus N <sub>1</sub> for Base-Line Engine and for Base-Line Engine Degraded by $1\%$ Increase in $\eta_c$ , $\eta_t$ , $\eta_{pt}$ , $\Lambda_5$ , and $\Lambda_N$	22
16	hppt Versus T7 for Base-Line Engine and for Base-Line Engine Degraded by $1\%$ Increase in $\eta_c$ , $\eta_t$ , $\eta_{pt}$ , $\Lambda_5$ and $\Lambda_N$	23
17	hppt Versus Wf for Base-Line Engine and for Base-Line Engine Degraded by 1% increase in $\eta_c$ , $\eta_t$ , $\eta_{pt}$ , $\Lambda_5$ , and $\Lambda_N$	24
18	Ambient Conditions Where Each Control Limit Determines Maximum Power of Base Engine	55
19	MPA System Configuration	94
20	MPA System Block Diagram	96
21	Vibrating Cylinder Pressure Transducer Schematic	104

#### LIST OF TABLES

Table		Page
I	Measurements at Low Power for Analytical Confirmation of Model Accuracy	26
II	Sensitivity Study - Effect of Sensor Errors at Low Power and at High Power on Power Remaining Accuracy Set I Sensors	28
Ш	Sensitivity Study - Effect of Sensor Errors at Low Power and at High Power on Power Remaining Accuracy Set II Sensors	31
IV	Error Summary Sheet - Prediction at 35.2% Horsepower at $N_2$ = 100% Using Constant "C" in Prediction Algorithm Set 1 Sensors	34
V	Error Summary Sheet - Prediction at 35.2% Horsepower at N $_2$ = 100% Using Interpolated "C" in Prediction Algorithm Set 1 Sensors	35
VI	Error Summary Sheet - Prediction at 19.7% Power at N $_2$ = 100% Using Interpolated "C" in Prediction Algorithm Set 1 Sensors	37
VII	Error Summary Sheet - Prediction at 35.2% Horsepower at $N_2$ = 100% Using Interpolated "C" in Prediction Algorithm Set II Sensors	38
VIII	Detailed Error List - Linearization of Nonlinear Differential Equations	41
IX	Detailed Error List - Effect of Nonstandard Day on the Value of "C" Base Engine	42
X	Detailed Error List - Prediction Error at Nonstandard Day Caused by Using Constant "C" ("C" Computed at Standard Day)	44
XI	Detailed Error List - Use of Base Engine "C" For Degraded Engine	e 45
XII	Detailed Error List - High Power Degradation Being Different from Low Power Degradation	1 46
XIII	Detailed Error List - Uncertainty in Compressor $W_{BL}$ and SPE at Maximum Power When Predicting at Low Power	47
XIV	Detailed Error List - Uncertainly in Compressor W <sub>BL</sub> and SPE at Low Power	48

Table		Page
xv	Steady-State Sensor and Control Accuracy Estimates	49
XVI	Detailed Error List - Sensor Errors at 35.2% Horsepower and at Maximum Horsepower Using Sensors at Low Power as Tabulated Best Available Sensors	50
XVII	Detailed Error List - Sensor Errors at 35.2% Horsepower and at Maximum Power Using Sensors at Low Power as Tabulated Standard Sensor Accuracy	51
XVIII	Detailed Error List - Sensor Errors at 19.7% Horsepower Using Sensors at Low Power as Tabulated Best Available Sensors	52
XIX	Detailed Error List - Sensor Errors at 35.2% Power and at Maximum Power Using Sensors at Low Power as Tabulated Best Available Sensors	53
xx	Detailed Error List - Sensor Errors at 35.2% Horsepower and at Maximum Horsepower Using Sensors at Low Power as Tabulated Standard Sensor Accuracy	54
XXI	Error Summary Sheet - Prediction at 35.2% Horsepower at N <sub>2</sub> = 100% With Feasible Improvements in Power Prediction Using Current "Best Available" Sensors Set 1 Low Power Sensors	57
XXII	Sensitivity Study - Effect of Sensor Error at Low Power and at High Power on Power Remaining Accuracy Set II Sensors	59
XXIII	Detailed Error List - Sensor Errors at 35.2% Horsepower Using Sensors at Low Power as Tabulated Best Available Sensors	62
XXIV	Error Summary Sheet - Prediction at 35.2% Power at $N_2$ = 100% Using Interpolated "C" in Prediction Algorithm Set III Sensors	63
XXV	Sensitivity Study - Effect of Sensor Errors at Low Power and at High Power on Power Remaining Accuracy Set IV Sensors	64
XXVI	Detailed Error List - Sensor Errors at 35.2% Power Using Sensors at Low Power as Tabulated Best Available Sensors	66
XXVII	Detailed Error List - Sensor Errors at 19.7% Horsepower Using Low Power Sensors as Tabulated Best Available Sensors	67

Table		Page
XXVIII	Error Summary Sheet Prediction at 35.2% Power at $N_2$ = 100% Using Interpolated "C" in Prediction Algorithm Set IV Sensors	68
XXIX	Detailed Error List - Sensor Errors at 57.5% Horsepower Using Low Power Sensors as Tabulated Best Available Sensors	70
xxx	Detailed Error List - Sensor Errors at 60.7% Horsepower Using Low Power Sensors as Tabulated Best Available Sensors	71
XXXI	Alternate Power Prediction Model Definition of Constants	76
XXXII	Summary of Errors for the Alternate Prediction System	77
XXXIII	Detailed Error List for the Alternate Prediction System Control Limit and Sensor Errors at High Power and Low Power	78
XXXIV	Detailed Error List for the Alternate Prediction System Errors Due to Degradation Since Last Calibration and Errors Due to Uncertainty in SPE and WBL	79
xxxv	Verification of Power Prediction Algorithm Using Actual Flight Test Data From AIDAPS Program Sensors Set II	88
XXXVI	Parametric Sensor Summary	101
XXXVII	Assumed Engine Tolerances	109

#### LIST OF SYMBOLS

A<sub>N</sub> Power turbine inlet nozzle effective area, in. <sup>2</sup>

A<sub>5</sub> Gas generator turbine inlet nozzle effective area, in. <sup>2</sup>

B(bij) Matrix relating relative variations in engine airflow pumping capacity, component efficiencies, and geometries to measured

relative variations in engine parameters sensed at low power

at a constant value of  $P_3/\delta_1$ .

C<sub>(Cij)</sub> Matrix relating the relative variations in maximum power at

each control limit to the computed relative variations in engine

pumping capacity, component efficiencies and geometries.

CF Nonoptimal N2 correction factor

DAN Computed relative change in A<sub>N</sub>

DA5 Computed relative change in A5

 $\delta_1$  Relative compressor inlet absolute pressure (=  $P_1/14.7$ )

DETAC Computed relative change in  $\eta_c$ 

DETAPT Computed relative change in  $\eta_{pt}$ 

DETAT Computed relative change in  $\eta_t$ 

DN1 Measured relative charge in  $N_1/\sqrt{\theta_1}$  at constant  $P_3/\delta_1$ 

DP7 Measured relative change in  $P_7/\delta_1$  at constant  $P_3/\delta_1$ 

DSHP Measured relative change in SHP/ $\delta_1 \sqrt{\theta_1}$  at constant P3/ $\delta_1$ 

DT3 Measured relative change in  $T_3/\theta_1$  at constant  $P_3/\delta_1$ 

DT7 Measured relative change in  $T_7/\theta_1$  at constant  $P_3/\delta_1$ 

DT9 Measured relative change in  $T_9/\theta_1$  at constant  $P_3/\delta_1$ 

DWA Computed relative change in  $W_a \sqrt{\theta_1}/\delta_1$ 

(engine pumping capacity)

DWF	Measured relative change in $W_f/\pmb{\delta}_1  \pmb{\theta}_1 \cdot {}^8$ at constant $P_3/\pmb{\delta}_1$
$\eta_{ m c}$	Compressor efficiency
$oldsymbol{\eta}_{ m pt}$	Power turbine efficiency
$\eta_{ m t}$	Gas generator turbine efficiency
$f_N$	Ambient temperature correction factor in computing maximum horsepower at $N_1\ limit$
$f_{T}$	Ambient temperature correction factor in computing maximum horsepower at $\mathbf{T}_7$ limit
$f_{\mathbf{W}}$	Ambient temperature correction factor in computing maximum horsepower at $\mathbf{W}_{f}$ limit
MPA	Maximum power available, hp
$N_1$	Gas generator turbine rotational speed, rpm
$N_2$	Free power turbine rotational speed, rpm
$P_{AM}$	Ambient pressure, psia
$P_l$	Compressor inlet total pressure, psia
$P_3$	Compressor discharge pressure, psia
P <sub>7</sub>	Power turbine inlet pressure, psia
SHP	Power turbine shaft horsepower, hp
$\mathtt{SHP}_{\mathbf{CO}}$	Optimal power turbine horsepower referred to sea-level standard-day conditions, hp
SHPON	Predicted maximum power at N <sub>1</sub> limit, hp
SHPOT	Predicted maximum power at T7 limit, hp
SHPOW	Predicted maximum power at Wf limit, hp
SHPOT	Predicted maximum power at N <sub>1</sub> limit, hp  Predicted maximum power at T <sub>7</sub> limit, hp

SHP <sub>REF</sub>	Sea-level standard-day value of horsepower at each control limit, hp
SPE/W <sub>a</sub>	Shaft power extraction/(lb/sec of compressor airflow), hp/(lb/sec)
ГАМ	Ambient temperature, °R
$T_1$	Compressor inlet total temperature, °R
T <sub>3</sub>	Compressor discharge total temperature, 'R
T <sub>7</sub>	Power turbine inlet total temperature, °R
Т9	Power turbine discharge total temperature, °R
$ au_{ m l}$	Relative compressor inlet absolute temperature (= $T_1/518.7$ )
$w_{BL}$	Compressor discharge bleed air as a percentage of total compressor airflow
$\mathbf{w_f}$	Engine fuel flow, pph
SUBSCRIPT	
В	Designates corrected base-line values
С	Designates quantities referred to sea-level standard-day conditions

#### INTRODUCTION

The purpose of this investigation was to determine the feasibility of developing a method to predict, with an accuracy of better than ±1%, the maximum power which can be produced by a helicopter gas turbine engine at full-power conditions. The prediction was to be made using information obtained from the engine while the engine was operated prior to lift-off at a partial-power condition of no more than 30% of normal rated power. The prediction method was to be capable of identifying the changes in maximum engine power available due to all possible types of engine deterioration and all ambient conditions. The study was based on a Lycoming T53-L13 gas turbine engine currently being used in the Army UH-l helicopter. The feasibility study was first based on the use of information available from existing sensors normally installed on the engine. It was determined, however, that the use of these sensors did not permit the desired ±1% accuracy to be obtained, and for this reason the feasibility study was expanded to include the use of additional and/or more accurate sensors. In addition, it was assumed that input information was available which provided atmospheric density data with an accuracy of ±.75%.

The following tasks were undertaken:

#### PHASE I - ESTABLISHMENT AND DEFINITION OF METHOD

- 1. A mathematical model of a method to predict MPA was constructed.
- 2. An estimated attainable MPA prediction accuracy was derived. Inaccuracies due to input information were identified.
- 3. Effects of power condition on MPA prediction accuracy were determined.
- 4. The hardware and components required to implement the method were described.

#### PHASE II - EVALUATION OF METHOD

- 1. An evaluation of the MPA prediction method was made using actual engine operational test data, in the following manner.
  - a. Input parameters were taken from engine operational test data at the recommended low-power condition.

- b. Using MPA prediction methods and the input parameters above, the MPA was computed.
- c. The actual maximum power, as taken from the engine test data used in (a) above, was then compared to the computed value determined in (b) above.
- 2. The above procedure was used to evaluate the ability of the prediction method to determine the maximum power available when the engine deteriorated in performance because of either internal deterioration or changes in atmospheric conditions.
  - a. In considering engine performance deterioration due to internal deterioration, the evaluation procedure used engine test data taken from engines known to be internally deteriorated, or from engines where internal deterioration was simulated.
  - b. In considering engine performance deterioration due to changes in atmospheric conditions, the evaluation procedure used data from at least three different engine test runs which included at least 29°F variation in ambient temperature, and from at least three different engine test runs which included at least 8,000 feet variation in altitude.

#### PHASE III - RECOMMENDATIONS

As a result of the feasibility investigation, recommendations were made in the areas of further sensor development, hardware implementation, system growth, and continued alternate MPA prediction system studies.

#### DISCUSSION

#### THE MATHEMATICAL SYSTEM MODEL

A mathematical system model was constructed that included the basic power prediction concept plus additional features for computing the errors in power prediction. A prerequisite for constructing this prediction model involved a detailed knowledge of the engine characteristics on which the maximum power was to be predicted.

A mathematical model of a "typical" Lycoming T53-L13 engine was evolved on an IBM 370 computer, based on engine modeling concepts developed by Hamilton Standard. The so-called "typical" T53-L13 engine was actually the average characteristics of test-cell data from 75 engines. This test data provided the steady-state values for N<sub>1</sub>, T<sub>3</sub>, P<sub>3</sub>, T<sub>9</sub>, SHP, and W<sub>f</sub> at standard-day conditions from idle to maximum power. The independent variables (such as component efficiencies and geometries) of the generic engine model were selected to duplicate the steady-state test data from the "typical" engine. The resulting computer model of the T53-L13 engine was then used to provide all required interrelationships. For example, changes in engine speed, temperature, pressure, or power resulting from changes in engine geometry, component efficiency, or air pumping capacity were computed. The engine model was used to compute the partial derivatives or influence of any engine p. rameter on any engine variable, called influence coefficients in this report.

The basic maximum power prediction concept is an extension of the engine diagnostic techniques previously developed by Hamilton Standard. This prediction concept is described as follows. The prediction computer contains three types of predetermined stored engine characteristics.

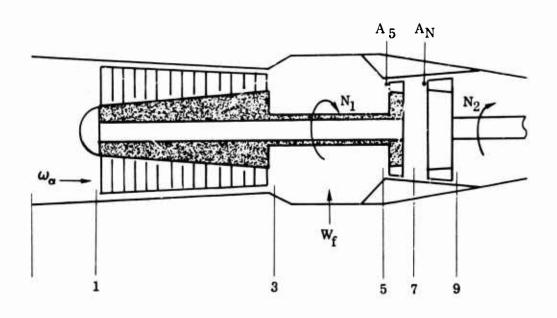
- 1. Base-line characteristics, consisting of steady-state locus of values for measured engine variables over the range of engine operation at which prediction can be performed.
- 2. A matrix of influence coefficients relating changes in the engine baseline variables to changes in such engine parameters as component efficiencies and geometries, called the "B-matrix" in this report.
- 3. A matrix of influence coefficients relating changes in engine parameters to changes in the engine power limit, called the "C-matrix" in this report.

Figure 1 shows a schematic diagram of a single-spool free-power turbine engine and identifies the various engine station numbers for future reference.

Measurements of engine variables and environmental conditions are obtained at low power. After measurement filtering and determination that the engine is in steady state, the measurement data is corrected to standard-day conditions. The percentage of point variations between the measurement data and stored base lines is determined. A gas-path analysis is then used to determine changes in engine characteristics (efficiencies, airflow, and geometries) from the values that existed when the stored base lines were determined. The percentage of point variations in engine characteristics, as determined at low power, is also assumed to exist at high power. Maximum power is determined by computing the fractional change in environment resulting from each change in engine characteristics and each change in environment from standard-day conditions. The complete computation of maximum power is performed for each control limit influencing maximum power, specifically a gas generator speed limit, a turbine discharge temperature limit, and a metered fuel limit. The lowest of the three computed values for maximum power is selected as the maximum available horsepower.

The gas-path analysis, performed on measurements obtained at low power, consists of a stored data matrix relating the increment between measurement and base-line data to variations in engine characteristics. This data matrix is determined in advance based on the engine thermodynamic relations. Numerical values of this matrix are stored at several low-power conditions, allowing a power prediction computation over a range of engine power. Similarly, the computation of the fractional change in maximum power is determined by a stored data matrix relating the variation of each engine characteristic and environment to the fractional change in maximum horsepower. This stored matrix is determined in advance based on engine thermodynamic relations at the standard-day maximum-power condition.

The basic tool in developing the power prediction algorithm is a technique developed by Hamilton Standard which quantitatively defines how the various engine performance parameters change with respect to each other or with changes in the environment or the engine fuel control. From a steady-state operating condition, a set of "influence coefficients" interrelating all the various engine performance parameters is determined. From this set of influence coefficients, the steady-state characteristics as well as the influence coefficients at any other power condition can be determined. The influence coefficients computed will ultimately be used in the power prediction scheme. Since the accuracy to which power can be predicted is affected by the accuracy of the influence coefficients, it is necessary that these coefficients be computed as precisely as possible.



Symbol	Definition
A <sub>5</sub>	Gas Generator Turbine Inlet Nozzle Effective Area
$A_N$	Power Turbine Inlet Nozzle Effective Area
N <sub>1</sub>	Gas Generator Turbine Rotational Speed
$N_2$	Free-Power Turbine Rotational Speed
$W_{\mathbf{f}}$	Engine Fuel Flow Rate
$oldsymbol{\eta}_{ m c}$	Compressor Efficiency
$\eta_{ m t}$	Gas Generator Turbine Efficiency
$oldsymbol{\eta}_{ m pt}$	Power Turbine Efficiency
$W_a$	Airflow Rate
WBL	Compressor Discharge Bleed Airflow
SPE	Compressor Shaft Power Extraction

Figure 1. Schematic Diagram of a Single-Spool Free-Power Turbine Engine.

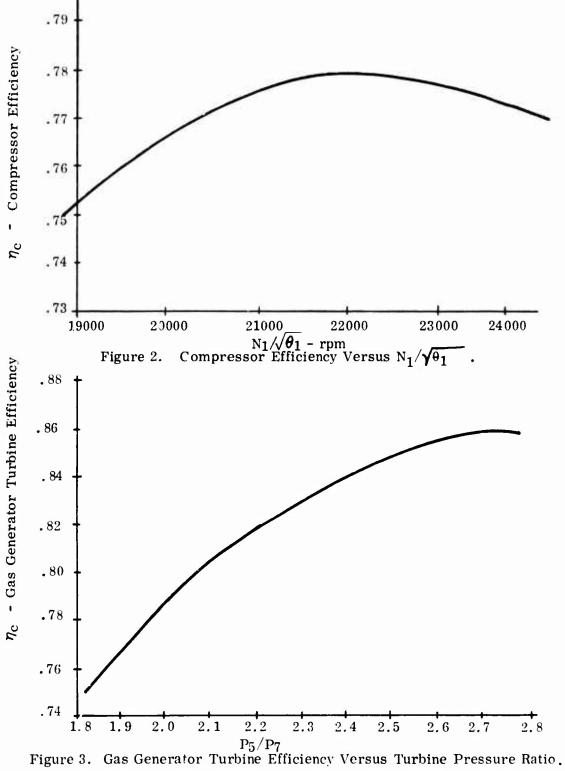
The accuracy of the influence coefficients is judged by determining if the computed steady-state characteristics correspond to actual steady-state characteristics of the engine obtained from tests. By varying compressor efficiency, gas generator turbine efficiency, and power turbine efficiency, the computed steady-state characteristics can be tailored to match actual steady-state characteristics for the particular engine.

Because of the large amount of test data available, the power prediction concept was specifically developed for the Lycoming T53-L13 engine. However, the general concepts are applicable to any free-turbine engine. A modern fuel control mode similar to the Hamilton Standard JFC 80 (used on the Lycoming LTC-4V-1 engine) was used for this program. In particular, it was assumed that maximum power is limited by a gas generator speed limit, a gas generator turbine discharge temperature limit, and a metered fuel flow limit. For component efficiencies as defined in Figure 2 through 4, the base-line steady-state characteristics as a function of corrected compressor discharge pressure are shown in Figures 5 through 10. Note that all engine characteristics are plotted as corrected quantities. Actual base-line data obtained by averaging test data of 75 T53-L13 engines are also shown in Figures 5 through 10.

Depending on the number and kind of sensor measurements made at low power, various specific cases of the basic power prediction concept were determined. (Selection of the specific cases most likely to yield the best power prediction was guided in part by the results of the engine diagnostic studies in the Army UH-1 AIDAPS Feasibility Program.) In this study, four such specific cases were evaluated. The four cases are defined by the following sets of low-power sensor measurements:

Set I 
$$(T_1, P_1, N_1, N_2, P_3, T_3, W_f, P_7, SHP, T_7)$$
  
Set II  $(T_1, P_1, N_1, N_2, P_3, T_3, W_f, SHP, T_9)$   
and assume  $\partial A_N/A_N = \partial \eta_{pt}/\eta_{pt}$   
Set III  $(T_1, P_1, N_1, N_2, P_3, T_3, W_f, P_7, T_9, T_7)$   
Set IV  $(T_1, P_1, N_1, N_2, P_3, T_3, W_f, SHP, T_7)$   
and assume  $\partial A_N/A_N = \partial \eta_{pt}/\eta_{pt}$ 

For the first set of sensors, the power prediction algorithm can be outlined in detail as follows:



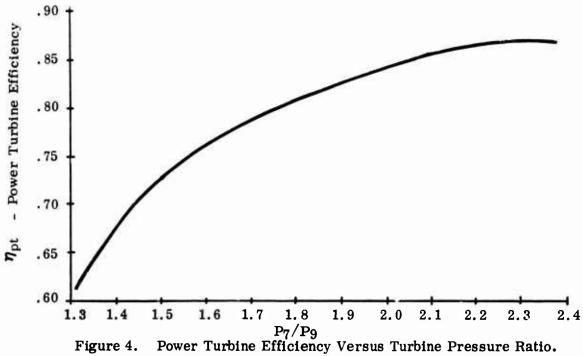


Figure 4.

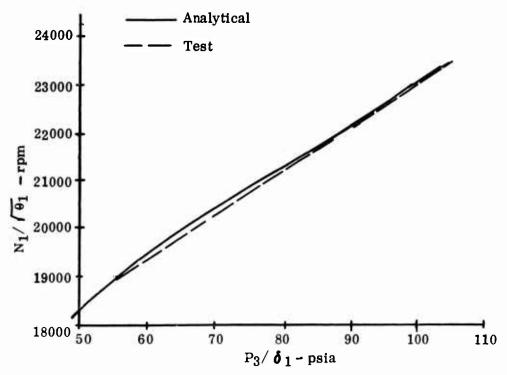


Figure 5. Corrected Gas Generator Speed Versus Corrected Compressor Discharge Pressure.

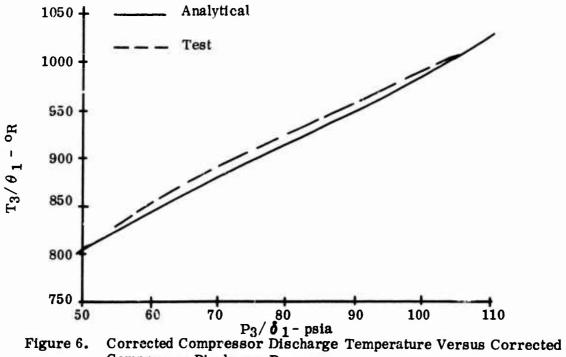


Figure 6. Compressor Discharge Pressure.

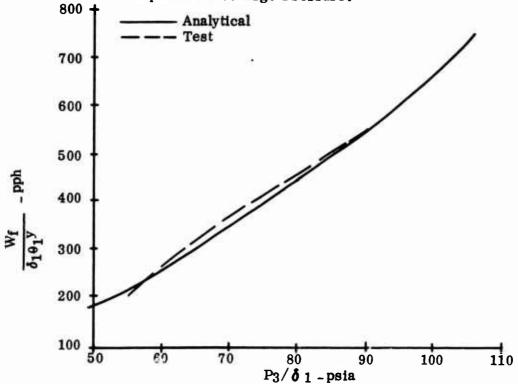


Figure 7. Corrected Fuel Flow Versus Corrected Compressor Discharge Pressure.

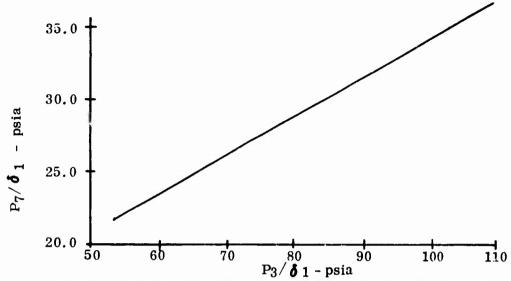


Figure 8. Corrected Gas Generator Turbine Discharge Pressure Versus Corrected Compressor Discharge Pressure.

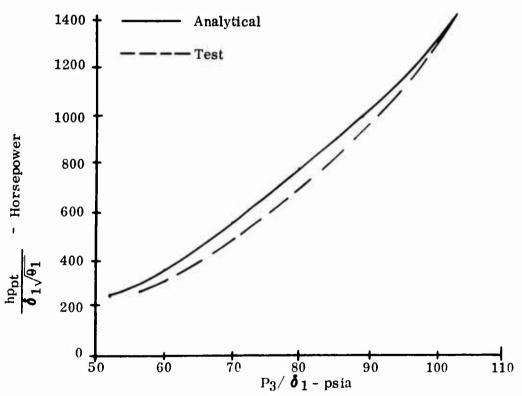


Figure 9. Corrected Power Turbine Horsepower Versus Corrected Compressor Discharge Pressure.

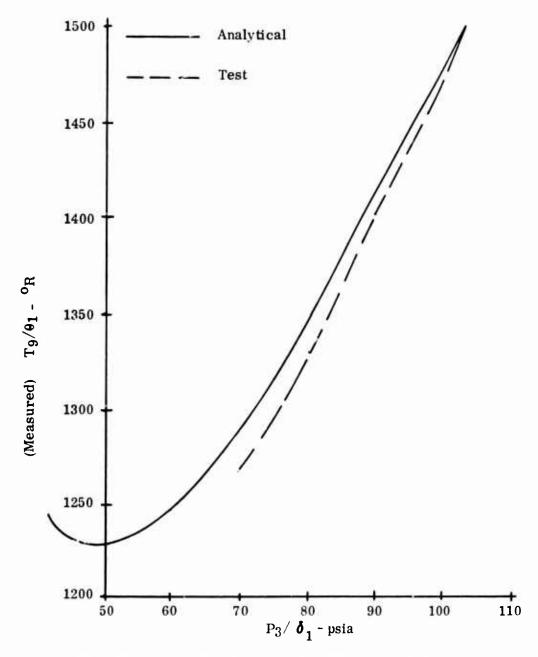


Figure 10. Corrected Power Turbine Discharge Temperature Versus Corrected Compressor Discharge Pressure.

I. Refer low-power sensor readings to standard-day conditions.

$$\theta_{1} = T_{1}/518.7$$

$$\delta_{1} = P_{1}/14.7$$

$$N_{1C} = N_{1}/\sqrt{\theta_{1}}$$

$$N_{2C} = N_{2}/\sqrt{\theta_{1}}$$

$$P_{3C} = P_{3}/\delta_{1}$$

$$T_{3C} = T_{3}/\theta_{1}$$

$$W_{fC} = W_{f}/\delta_{1}\theta_{1}^{Y}$$

$$SHP_{C} = SHP/\delta_{1}/\overline{\theta_{1}}$$

$$P_{7C} = P_{7}/\delta_{1}$$

$$T_{7C} = T_{7}/\delta_{1}$$

II. Determine base-line values from stored hardware at the same  $P_{\rm 3C}$  value as in I.

$$N_{1CB} = f_{1}(P_{3C})$$
 $N_{2CB} = f_{2}(P_{3C})$ 
 $T_{3CB} = f_{3}(P_{3C})$ 
 $W_{fCB} = f_{4}(P_{3C})$ 
 $SHP_{CB} = f_{5}(P_{3C})$ 
 $P_{7CB} = f_{6}(P_{3C})$ 
 $T_{7CB} = f_{7}(P_{3C})$ 

III. Compute optimal corrected SHP from corrected SHP.

Correction Factor (CF) = 
$$\left[ (N_{2C} - N_{2CB})/N_{2CB} \right]^2$$
 (3)

$$SHP_{CO} = SHP_{C}/(1 - CF)$$
 (4)

IV. Compute the relative deviation of the measurement data from the stored base-line data.

$$DN1 = (N_{1}C - N_{1}CB)/N_{1}CB$$

$$DT5 = (T_{3}C - T_{3}CB)/T_{3}CB$$

$$DWF = (W_{f}C - W_{f}CB)/W_{f}CB$$

$$DSHP = (SHP_{CO} - SHP_{CB})/SHP_{CB}$$

$$DT7 = (T_{7}C - T_{7}CB)/T_{7}CB$$

$$DP7 = (P_{7}C - P_{7}CB)/P_{7}CB$$
(5)

V. Compute the variations in airflow pumping capacity, efficiencies, and geometries from the following matrix equation.

$$\begin{bmatrix} DWA \\ DETAC \\ DETAT \\ DETAPT \\ DAS \\ DAN \end{bmatrix} = \begin{pmatrix} (\eta_{C} - \eta_{CB})/\eta_{CB} \\ (\eta_{t} - \eta_{tB})/\eta_{tB} \\ (\eta_{pt} - \eta_{ptB})/\eta_{ptB} \\ (A_{5} - A_{5B})/A_{5B} \\ (A_{N} - A_{NB})/A_{NB} \end{bmatrix} = B \begin{bmatrix} DN1 \\ DT3 \\ DWF \\ DNPPT \\ DT7 \\ DP7 \end{bmatrix} (6)$$

The B-matrix in Equation 6 is computed from the influence coefficients and is a function of the set of engine parameters measured at low power. A different set of low-power sensors would require a different B-matrix.

Since the value of P<sub>3C</sub> at which the low-power measurements are determined is not known in advance, the B'-matrix has been determined and stored at several values of P<sub>3C</sub>, and linear interpolation on P<sub>3C</sub> is used to obtain the actual B'-matrix to be used in Equation 6. In the sensitivity studies to be discussed later, an augmented B'-matrix was used to take into account the effects of uncertainty in compressor discharge air bleed and shaft power extraction at low power on predicted power.

VI. Compute maximum power at each of the control limits from the following equation.

On the T7 temperature limit

SHPOT = 
$$(SHP_{REF}) \delta_1 (1 + DWA)^{C11} (1 + DETAC)^{C12} (1 + PETAT)^{C13}$$
  
 $(1 + DETAPT)^{C14} (1 + DA5)^{C15} (1 + DAN)^{C16} f_T (T_{AM})$  (7)

On N<sub>1</sub> speed limit

SHPON = (SHPREF) 
$$\delta_1^{C27}$$
 (1 + DWA)<sup>C21</sup> (1 + DETAC)<sup>C22</sup> (1 + DETAT)<sup>C23</sup>  
(1 + DETAPT)<sup>C24</sup> (1 + DA5)<sup>C25</sup> (1 + DAN)<sup>C26</sup>  $f_N$  (T<sub>AM</sub>) (8)

On Wf limit

SHPOW = 
$$(SHP_{REF}) \delta_1^{C37} (1 + DWA)^{C31} (1 + DETAC)^{C32} (1 + DETAT)^{C33}$$
  
 $(1 + DETAPT)^{C24} (1 + DA5)^{C25} (1 + DAN)^{C26} f_W (T_{AM})$  (9)

In Equations 7 through 9 it has tacitly been assumed that the changes in component efficiencies, airflow, and geometries computed at low power also apply at high power. This assumption is a source of power prediction error which is taken into account in the error analysis. In the sensitivity studies to be discussed later, Equations 7 through 9 were expanded to take into account the effects of control and sensor inaccuracy at high power on predicted power as were the effects of uncertainty in compressor discharge air bleed and shaft power extraction and high power on predicted power. These factors are important in evaluating the overall accuracy of the model; however, they do not enter into the actual power prediction algorithm.

From the horsepowers computed at the three control limits, the minimum is chosen as the maximum power available; that is,

$$MPA = MIN (SHPOT, SHPON, SHPOW)$$
 (10)

For any other set of low-power measurements, the procedure for determining the maximum power available is analogous to the procedure described above except that a different B-matrix is used. The "C coefficients in Equations 7 through 9 do not change unless the engine control mode is changed.

Of the many variables which enter into the power prediction scheme, the one which has the widest range of variation is ambient temperature. The ambient temperature range of -60°F to 120°F represents a -42.3% to +11.8% variation from the standard-day ambient temperature value of 518.7°R ( $\approx$ 59°F). In the development of the power prediction algorithm, the effect of ambient temperature appeared as a factor in the equation for horsepower of the form [  $T_{AM}/(T_{AM})_{REF}$ ] CTAM. In using this factor, it was found that over the wide ambient temperature range expected, the exponent CTAM varied sufficiently such that using a constant value for CTAM resulted in prohibitively large errors in predicted horsepower.

An alternate procedure was chosen which uses an ambient temperature correction factor to account for ambient temperature variations. At a given ambient temperature, the ambient temperature correction factor is defined to be the ratio of actual horsepower at the specified ambient temperature to the actual horsepower at the standard-day reference temperature of 518.7°F. A separate ambient temperature correction factor is required for each of the three control limits and is shown in Figures 11, 12 and 13 for the base-line engine.

As the engine degrades, the ambient temperature correction factor will shift. The power prediction algorithm does not take into account this shift but uses the ambient temperature correction factors defined in Figures 11, 12 and 13 for all engines. As a result, an error in the predicted power will be introduced for degraded engines for nonstandard-day ambient temperatures.

Engine diagnostic studies in the UH-l AIDAPS feasibility program have demonstrated the need for determining that the engine is sufficiently near steady state to obtain meaningful steady-state data. This evaluation has not been duplicated in this feasibility study, as the evaluation in the AIDAPS program is applicable to power prediction. Therefore, the technique for determining steady-state

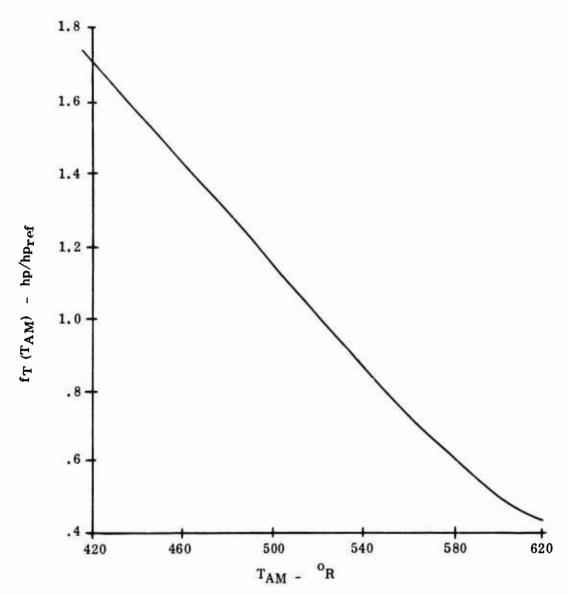


Figure 11.  $T_7$  Limit Mode Ambient Temperature Correction Factor Versus  $T_{\mbox{AM}}$ .

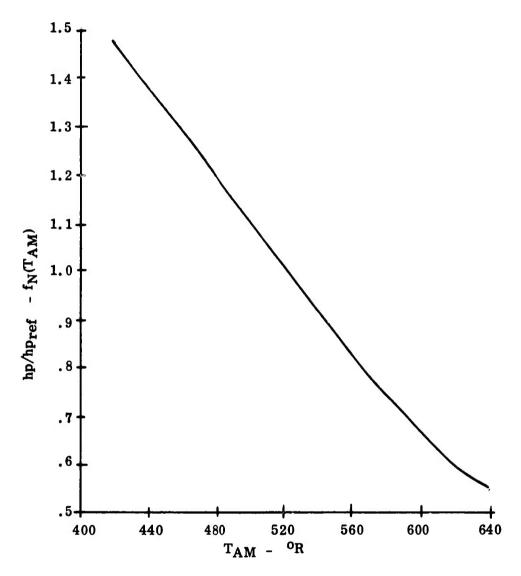


Figure 12.  $N_1$  Limiting Mode Ambient Temperature Correction Factor Versus  $T_{A\,M}$ .

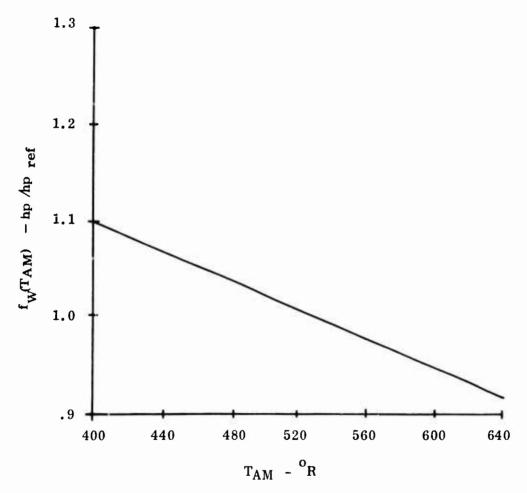


Figure 13.  $W_f$  Limit Mode Ambient Temperature Correction Factor Versus  $T_{AM}$ .

conditions evolved in the UH-1 AIDAPS program will also be used in the power prediction algorithm. Satisfying all three of the following conditions indicates that the engine is sufficiently near steady state to take measurements for prediction.

- 1.  $N_{1 \text{ MAX}} N_{1 \text{ MIN}}$  within a frame  $\leq 200 \text{ rpm}$
- 2.  $|\Delta N_{1C}|$  between frames  $\leq 100$  rpm
- 3.  $T_{T7C~MAX} T_{T7C~MIN} \le 2.5^{\circ}F$  for 60 seconds

One frame of data represents two seconds of measurement time. There are four speed measurements and four temperature measurements in one frame.

The power prediction algorithm involves the use of predetermined B'-matrix, "C'-matrix, and base lines. As noted, the B'-matrix relates the variations in engine efficiency, geometry, and airflow pumping capacity to measured variations in speed, temperature, pressure, power, and fuel at low power, whereas the C'-matrix relates the variations in maximum power on each control limit to the computed variations in maximum power on each control limit to the computed variations in engine efficiency, geometry, and airflow pumping capacity. Studies have indicated that neglect of the changes in the B'-matrix and C'-matrix resulting from engine degradation causes a relatively minor error in the power prediction algorithm. Therefore, the B'-matrix and C'-matrix, as computed for the typical engine, may be used in all units; i.e., every prediction computer will contain identical stored B' and C'-matrices for the same engine model.

However, a review of the actual steady-state data from the 75 engines tested shows relatively large engine-to-engine variations in the base lines. As a result, it appears necessary to measure and store a unique set of base-line values for each engine to avoid significant errors in prediction. This requires an initial steady-state "calibration" for each engine to be loaded into the prediction computer. The power prediction can then accurately predict the effect of any subsequent engine degradation. It should be noted that this initial calibration requirement of the engine base lines is based on observed engine-to-engine variations of the T53-L13 model and may not apply to another engine model.

#### CONFIRMATION OF MODEL ACCURACY

To analytically evaluate the validity of the algorithm for computing MPA, the base-line engine was "degraded" by making a 1% increase in  $\eta_C$ ,  $\eta_t$ ,  $\eta_{pt}$ ,  $\Lambda_{5}$ , and  $\Lambda_{N}$ . The influence coefficients were used to obtain a set of steady-state engine data for the "degraded" engine. This data served as the source of low-power measurements for the degraded engine. The MPA was calculated using the steady-state characteristics for the degraded engine and using the power prediction algorithm. All computations were made at sea-level standard-day conditions and at optimal  $N_2$  speed.

In order to determine MPA when on  $N_1$  speed limiting for the steady-state characteristics of the degraded engine, it is first necessary to determine the new  $N_1$  limiting speed. It is assumed that the fuel control is a  $W_f/\delta_1$  type droop control with a droop slope given by:

$$\left(\frac{\partial W_{f}/\delta_{1}}{W_{f}/\delta_{1}} / \frac{\partial N_{1}/\sqrt{\theta_{1}}}{N_{1}/\sqrt{\theta_{1}}}\right) = -6.5$$
Droop Line

For sea-level standard-day conditions, Figure 14 shows a plot of  $W_f/\delta_1$  vs.  $N_1/\sqrt{\theta_1}$  for both the base-line and degraded engines. Also shown in Figure 14 is the control droop line which passes through the base-line point of  $N_1/\sqrt{\theta}$  = 24, 700 rpm and Wf/ $\delta_1$  = 793 pph and whose slope is given by Equation 11. The intersection of the "degraded" steady-state line with the droop line defines the N<sub>1</sub> limiting speed for the degraded engine. From Figure 14, the limiting speed for the degraded engine is approximately 24,820 rpm. Figure 15 shows power turbine horsepower as a function of gas producer speed for the base-line and degraded engines. The horsepower at the N<sub>1</sub> limiting speed for the degraded engine is designated by the symbol "SHPON" and is found to be 1383 hp. Figure 16 shows power turbine horsepower as a function of T7 for the base-line and degraded engines. At the T<sub>7</sub> limit value of 1743°R, the horsepower for the degraded engine designated by "SHPOT" is found to be 1545 hp. Figure 17 shows power turbine horsepower as a function of fuel flow for the base-line and degraded engines. At the Wf limit value of 793 pph, the horsepower for the degraded engine designated by SHPON is found to be 1440. The above results were analytically determined from the steady-state characteristics of the "degraded" engine. From these results, the MPA is given by:

$$MPA = MIN (SHPON, SHPOT, SHPOW) = 1383 hp$$
 (12)

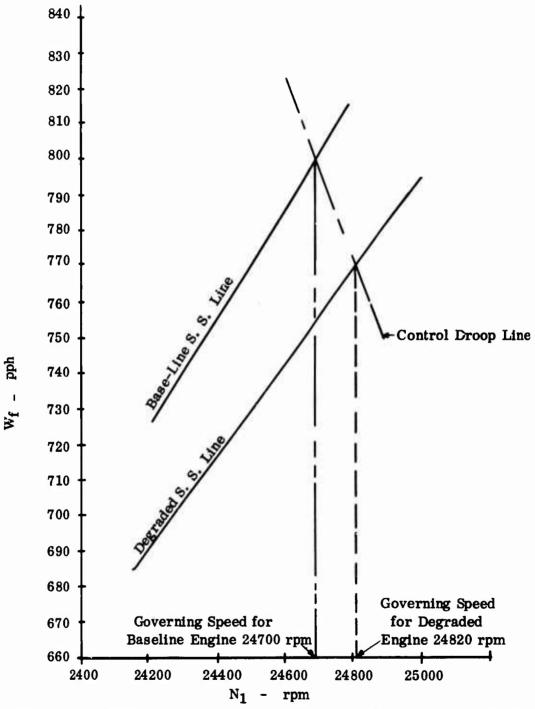


Figure 14. Wf Versus N<sub>1</sub> for Base-Line Engine and for Base-Line Engine Degraded by 1% Increase in  $\eta_c$ ,  $\eta_t$ ,  $\eta_{pt}$ , A<sub>5</sub> and A<sub>N</sub>.

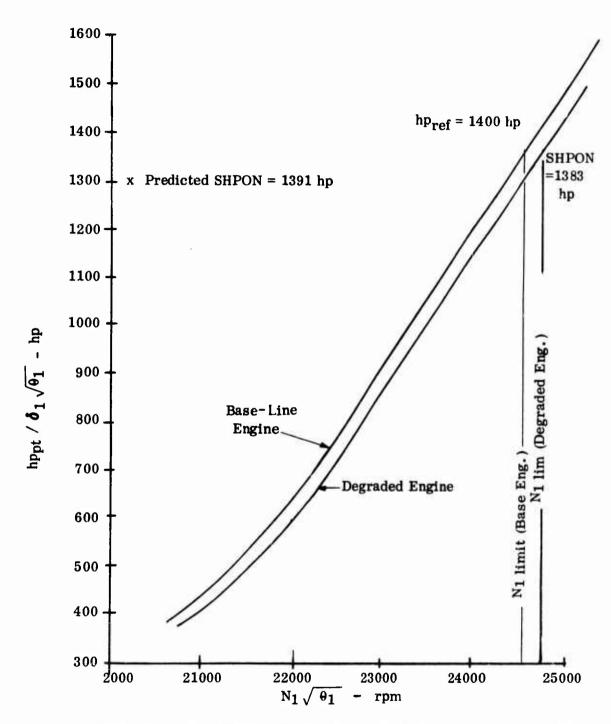


Figure 15. hp  $_{pt}$  Versus N1 for Base-Line Engine and for Base-Line Engine Degraded by 1% Increase in  $~\eta_c,~\eta_t,~\eta_{pt},~A_5,~$  and  $A_N$  .

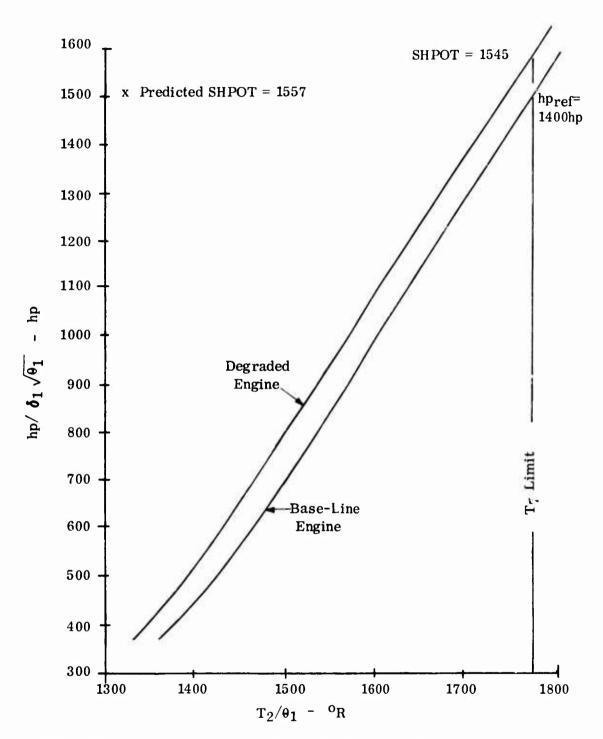


Figure 16. hp  $_{pt}$  Versus T  $_7$  for Base-Line Engine and for Base-Line Engine Degraded by 1% Increase in  $\pmb{\eta}_{\,c},\,\pmb{\eta}_{\,t}\,\pmb{\eta}_{pt},\,\mathbf{A}_5$  and  $\mathbf{A}_N$  .

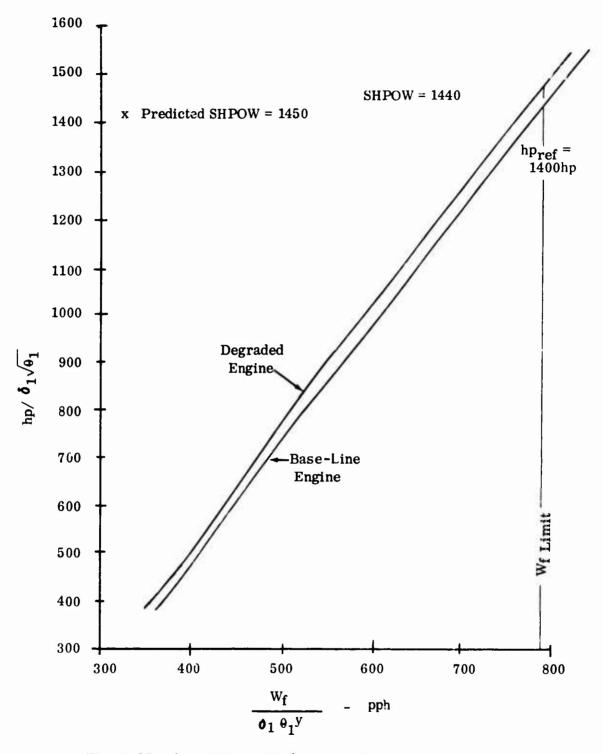


Figure 17. hp  $_{pt}$  Versus W  $_f$  for Base-Line Engine and for Base-Line Engine Degraded by 1% Increase in  $\eta_c,~\eta_t,~\eta_{pt},~A_5,~\&~A_N$  .

At a relatively low power condition, input data for the power prediction program was obtained from the steady-state characteristics of the degraded engine at a constant  $P_3/\delta_1$  value. Table I shows the steady-state values used as measurements in the power prediction program at  $P_3/\delta_1 = 60.3647$  psia.

The results from the power prediction program are summarized as follows:

SHPON = 1390.9448 hp (N<sub>1</sub> Limiting)

SHPOT = 1557.09668 hp (T<sub>7</sub> Limiting)

SHPOW = 1449.7737 hp (Wf Limiting)

These predicted values are also shown in Figures 15, 16, and 17. From the power prediction program, the MPA is given by

MPA = MIN (SHPON, SHPOT, SHPOW) = 1377,877 hp

The results from the power prediction program are within 1% of the analytical results based on the steady-state characteristics. The small errors are due in part to errors in precisely determining the steady-state characteristics of the degraded engine, errors in predicting power at each control limit, and computer round-off errors. The results, however, do show that the power prediction algorithm is conceptually correct.

## PARAMETRIC SENSITIVITY STUDIES AND ERROR ANALYSES

The following conditions were assumed for the parametric sensitivity studies and error analyses:

- 1. The pilot-selected N2 speed and collective pitch result in 100% rotor speed at the maximum power condition being predicted.
- The detailed error analysis was performed using the T53-L13 helicopter engine with a modern helicopter engine control similar to the control mode used on the Lycoming LTC-4V-1 engine. Specifically, it is assumed that maximum power is limited by the least of three control limits on N₁, T7 and Wf.

TABLE I. MEASUREMENTS AT LOW POWER FOR ANALYTICAL CONFIRMATION OF MODEL ACCURACY

 $P_3/\delta_1 = 60.3647$  psia

Parameter	Base-Line Values	Degraded Engine*
$N_1/\sqrt{\theta_1}$	20, 700 rpm	20, 800 rpm
$\tau_3/\theta_1$	851.456°R	847.824°R
$\mathbf{w_f}/(\boldsymbol{\delta_1}\boldsymbol{\theta}_1^{\mathbf{y}})$	354.036 pph	343.846 pph
P7/61	23.7552 psia	23.7057 psia
$(\mathrm{hp}_{\mathrm{pt}})/(\pmb{\delta}_1\sqrt{\pmb{\theta}_1})$	369.493 hp	363.159 hp
$T_7/\theta_1$	1369.86°R	1333.32°R

<sup>\*</sup> Measurements corresponding to 1% increase in  $\eta_C,~\eta_t,~\eta_{pt},~\Lambda_5,~$  and  $A_N.$ 

- 3. Specific assumptions needed to perform the error analyses were included in Appendix I.
- 4. The interest in power prediction is limited to lift-off conditions; therefore, effects of flight speed need not be included in the error analysis.

Sensitivity studies were conducted to determine the influence of each pertinent factor on the accuracy of computed maximum power. Sensor errors at low power and at high power are considered separately.

The results of the sensitivity study for Set I sensors are shown in Table II. Table II also gives the error in predicted power due to uncertainty in compressor bleed air (W<sub>BL</sub>) and compressor shaft power extraction (SPE/W<sub>a</sub>) at low power and high power. As an example, Table II shows that an N<sub>1</sub> speed sensor indicating a speed 1% higher than actual speed at low power (N<sub>1</sub> = 21, 900 rpm, SHP = 35.2% of maximum power) causes the computed maximum power to be 4.02% less than the actual maximum power if the engine is on the T<sub>7</sub> control limit.

Note that  $N_2$  speed at both low and high power is defined as 100%. The  $N_2$  speed is actually dependent on both pilot-selected  $N_2$  limiting speed and pilot-selected collective pitch. Similarly, it is assumed that  $N_2$  speed is 100%, and collective pitch is selected to cause the engine to encounter a control limit while at high power.

The percent-of-point temperature and pressure sensor errors are percentage of absolute temperature (°R) and absolute pressure (psia), respectively. (WBL) is percentage of compressor airflow. (SPE/ $W_a$ ) is horsepower per pound per second of compressor airflow.

Table III gives the sensitivity table for Set II sensors. Note that this set of sensors is not sufficient to uniquely determine the six performance parameters. A relationship between the independent variables must be assumed. This assumed relationship (although inexact) is that  $\partial A_N/A_N = -\partial \eta_{pt}/\eta_{pt}$  and was demonstrated to be a good approximation in the AIDAPS engine diagnostic program. Note also that this set of measurements at low power was obtained on actual engines in the AIDAPS program.

The total horsepower prediction errors for prediction algorithms using the first two sets of sensors are summarized in Tables IV, V, VI, and VII. Table IV is an error summary when using Set I sensors. Using constant "C" coefficients

TABLE II.		NSITIVITY STUDY D AT HIGH POWE T I SENSORS	DY - EFFEC WER ON POW	SENSITIVITY STUDY - EFFECT OF SENSOR ERRORS AT LAND AT HIGH POWER ON POWER REMAINING ACCURACY SET I SENSORS	R ERRORS AT	- EFFECT OF SENSOR ERRORS AT LOW POWER R ON POWER REMAINING ACCURACY	~
% Max Power % of Point	Limit Mode	$N_1 = 21,700$ hp = 35.2%	$N_1=21,200$ hp = 27.%	$N_1 = 20, 700$ hp = 19, 7%	$N_1 = 20, 200$ hp = 13. 6%	$N_1 = 19, 700$ hp=8.6%	$N_1 = 24,700$ hp = 100%
θhp ΘN <sub>1</sub>	$N_1$ $T_7$	-7.45 -4.02	-8.50 -5.41	-10.08 - 7.32 - 5.83	-12.58 -10.27	-15.14 -13.23 -11.57	-3.71
θ h <b>p</b> θ T3	$^{N}_{1}$ $^{1}_{7}$ $^{N}_{7}$	14	19	24	. 29	90.	
∂hp ∂w <sub>f</sub>	$^{ m N}_{ m 1}$	59 + .01 -1.46	53 + . 02 -1. 46	60 + .03	60 + .03 -1.46	61 + .04 -1.46	+ .56 0. +1.47
θh <b>p</b> /θ SHP	$N_1$ , $T_7$ , $W_f$	1.00	1.00	1,00	1.00	1.00	1

		I	TABLE II. C	Continued			
% Max Power % of Point	Limit Mode	$N_1 = 21,700$ hp = 35.2%	$N_1 = 21, 200$ hp = 27. %	N <sub>1</sub> =20, 700 hp= 19. 7%	$N_1 = 20, 200$ hp = 13.6%	$N_1 = 19, 700$ hp = 8.6%	N <sub>1</sub> =24,700 hp=100%
$\frac{\partial h\mathbf{p}}{\partial \mathbf{T}_7}$	$^{N}_{1}$	20	25	30	34	.37	0.
	7 J Wf	-4.40	-4.46	-4.5 <b>2</b> - 0.7	-4.58	-4.64	-4.31
9hp	$^{\rm N}_1$	- 39	49	- , 59	67	- ,75	0.
$\partial P_7$	$^{\mathrm{T}_{7}}$ Wf	71	86 69.	<b>-1.2</b> 3	-2.30	-1.98	0 0
∂hp/∂N <sub>2</sub>	$N_1$ $T_7$ , $W_{\hat{f}}$	+1.03	+1.44	+2.10	+3.25	+5.37	1
θhp θ Τ <sub>1</sub>	$N_1$ $T_7$	+3,68 +5.36	+4.20	+4.97	+6.20	+7.03	.76
	$\mathbf{w}_{\mathbf{f}}$	+1.79	+2. 29	+3.02	+4.29	+5.15	0
$\frac{\partial hp}{\partial P_1}$	$^{N}_{1}$	45	41	90	<b>32</b> + . 30	25	
	Wſ	+ .84	+ .94	+1.12	+1.32	+1.57	37

		T/T	TABLE II. Co	Continued			
$ec{ec{arphi}}_{o}^{n}$ Max Power $ec{ec{arphi}}_{o}^{n}$ of Point	Limit Mode	$N_1 = 21,700$ hp = 35,2%	$N_1$ =21, 200 hp=27. %	$N_1 = 20, 700$ hp = 19, 7%	$N_1 = 20, 200$ hp=13.6%	$N_1 = 19,700$ hp= 8.6%	$N_1 = 24,700$ hp = $100\%$
dye	Z I						0.
$^{ ho}$ P <sub>3</sub>	${ m T_7}$ ${ m w_f}$	+ .14	+ .05	90. +	+ + +	+ .03	· ·
∂hp/∂N <sub>1</sub> Set	$^{N}_{1}$	.0	0.	0.	0.	0.	+3.71
∂hp/∂ T <sub>7</sub> Set	$T_{7}$	0.	0.	0.	.0	•	+4.31
9hp	$_{1}^{N}$	+ .70	89•+	99°+	06. +	+1.11	82. +
G W BL	<i>L</i> .I.	+3.24	+3.25	+3.26	+2.78	+2.35	+3. 29
	$\mathbf{w_f}$	+1.46	+1.38	+1.37	+1.19	+1.04	+1.43
* dhe	$^{N}_{1}$	39	43	48	53	58	+ .22
$\left[ \frac{3PE}{WA} \right]$	$T_{7}$	+2.26	+2.44	<b>+2</b> .60	+2.82	+3.02	-1.57
:	$\mathbf{w}_{\mathbf{f}}$	+ .95	+1.02	+1.07	+1.14	+1.21	89
Notes	*Units are	*Units are 7 Max Power/(hp/rps		of Airflow)			
	$N_2 = 100\%$	J.W.					
	Standard [	Day Conditions					

	1 8 1												
	$N_1 = 24.700$ hp=100%	-3.71	0.0	0.0	0.0	0.0	0.0	+ .56	0.0	+1.47	1	11	1
OW POWER	$N_1$ =19, 700 hp=8, 6%	-12.63	- 5.71	- 8.32	. 28	+ .23	+ 0.00	. 61	+ .32	- 1.50	08. +	+ .17	+ .63
RORS AT LO	$N_1 = 20, 200$ hp=13.6%	-10.85	5, 39	6.48	. 24	+ .16	02	19	+ .36	- 1.50	08.	+ .21	+ .67
SENSOR ER REMAINING	$N_1 = 20,700$ hp=19.7%	-9.03	-4.57	-4.54	20	*0. +	04	09	+ .36	-1.49	+ .81	+ .23	+ .72
SENSITIVITY STUDY - EFFECT OF SENSOR ERRORS AT LOW POWER AND AT HIGH POWER ON POWER REMAINING ACCURACY SET II SENSORS	N <sub>1</sub> ≕21,200 hp≂27 <i>%</i>	98.7-	3.82	-3.31	16	+ .04	05	09	+ .42	1.49	+ .82	+ .25	+ .78
TY STUDY - GH POWER VSORS	N <sub>1</sub> =21, 700 hp=35. 2%	7.08	-3.14	-2.44	19	01	90	09	+ .54	-1.49	98. +	+ .23	98. +
SENSITIVI AND AT HI SET II SEI	$N_1 = 22, 800$ hp=57. $5\%$	5, 55	-1.74	-1.14	05	05	90	59	89. +	-1.48	+ .92	+ .18	+ .92
,Е Ш.	Limit Mode	z	$^{7}$	Wf	N 1	$^{\mathrm{T}}_{7}$	Wf	Z 1	$_7^{\rm T}$	Wf	$\mathbf{z_{1}}$	$T_7$	Wf
TABLE III.	T Max Power Tof Point		7.		d h b	3		d h e					

			TABE	TABLE III. Con	Continued			
# Max Power % of Point	Limit Mode	$N_1 = 22,800$ hp=57.5%	$N_1 = 21,700$ hp=35.2%	$N_1 = 21, 200$ hp=27%	$N_1 = 20, 700$ hp=19.7%	$N_1 = 20, 200$ hp=13.6%	$N_1 = 19,700$ hp=8.6%	$N_1 = 24,700$ hp=100%
9hp	z 1	04	90	80.	10	13	.15	0.0
610	$_{7}^{\mathrm{T}}$	-4.56	-4.52	-4.48	-4.46	-4.40	-4.36	-4.31
	W	60°+	+ .17	+ .20	+ .23	+ .26	+ .28	0.0
9hp	Z	+ .40	88.	+1.22	+1.75	+2.71	+4.48	-
<b>N</b> N 0	$^{7}$	+ 38	+ .73	+ .92	+1.14	+1.41	+1.78	-
	Wf	8e. +	+ .84	+1.14	+1.61	+2.41	+3.87	1
9hp	z	+2.75	+3.52	+3.94	+4,56	+5, 59	+6.28	92
1,0	$^{L_{7}}$	+4.64	+5,15	+5,35	+5,58	+5.72	+5,42	0.0
	$M_{\mathbf{f}}$	+1.02	+1.46	+1,79	+2.30	+3,16	+3.70	0.0
gh6	N 1	81	89	02	92	92	78	0.0
1	$^{\mathrm{L}}$	-1.05	16	98	87	71	09	0.0
	$M_{\mathbf{f}}$	+ .38	+ .51	+ .52	+ .54	09. +	+ .65	37
9hp	N	+ .33	+ .22	+ .20	+ .22	+ .24	+ .24	0.0
053	$^{\mathrm{L}_{2}}$	+ .16	+ .10	90. +	+ .05	+ .03	+ .02	0.0
	Wf	+ .23	+ .08	+ .07	+ .07	90°+	+ .03	0.0

			TAB	TABLE III. Co	Continued			
% Max Power % of Point	Limit Mode	$N_1 = 22,800$ hp=57.5%	$N_1$ =21, 700 hp=35. 2%	$N_1 = 21, 200$ hp=27%	$N_1$ =21, 200 $N_1$ =20, 700 hp=27% hp=19.7%	$N_1 = 20, 200$ hp=13.6%	$ m N_1$ =20, 200 $ m N_1$ =19, 700 $ m N_1$ =24, 700 hp=13.6% hp=8,6% hp=100%	$N_1 = 24,700$ hp=100%
θhp/θ N <sub>1</sub> SET	Z	0.	0.	0.	0.	0.	0.	+3.71
θ hp/θ T <sub>7</sub> SET	$T_7$	0	.0	0.	0.	.0	0.	+4.31
θhp θWBL	$r_1^N$	+ .74	+ .72	+ .70	+ .69	+ .67	+ .65	78
	Wf	+1.44	+1.43	+1,46	+1.42	+1,41	+1,41	-1.43
- Land	$^{N}_{1}$	315	390	425	463	514	560	+ .22
( acare	Wf	+0.84	696°+	+1.04	+1,10	+1,18	+1.25	89
Notes:	* Uni (hp,	Units are % Max Power (hp/pps of airflow) $N_2 = 100\%$ Standard Day Conditions	x Power ow)				}	

TABLE IV. ERROR SUMMA AT N2 = 100% U SET I SENSORS	ERROR SUMMARY SHEET - PREDICTION AT 35.2% HORSEPOWER AT N2 = 100% USING CONSTANT "C" IN PREDICTION ALGORITHM SET I SENSORS	EET - PR	EDICTION	AT 35.2% HOREDICTION	ORSE POWER ALGORITHM
Error Source	N <sub>1</sub> Limit % Power	T <sub>7</sub> Limit W <sub>f</sub> Limit % Power % Power	W <sub>f</sub> Limit % Power	Detailed Error List	Applicable Notes (Appendix I)
Linearization of nonlinear differential equations	0.50	1.23	0.35	νш	1.
Nonstandard day	0.31	4.98	3.61	IX, X	2,3
Use of base engine "C" for degraded engine	0.03	0.50	0.03	XI	4,5
High-power degradation being different from low-power degradation	0.22	0.68	0.22	XII	9
Change in actual power due to uncertainty in WBL and SPE at high power	1.17	4.96	2.15	хш, п	<b>-</b>
Uncertainty in WBL and SPE low power	0.37	1.76	0.78	XIX	œ
Sensor errors at low power	2.61	3.60	2.33	xv, xvi, II	6
Control limit and sensor errors at high power	0.52	1.55	0.74	хv, хvı, п	ര
Total Error:	3.00	8.37	4.94		

TABLE V. ERROR AT N2 = ALGOR	ERROR SUMMARY SHEET - AT N <sub>2</sub> = 100% USING INTERP ALGORITHM SET I SENSORS	EET - PRE NTERPOLA INSORS	DICTION A	ERROR SUMMARY SHEET - PREDICTION AT 35.2% HORSEPOWER AT N <sub>2</sub> = 100% USING INTERPOLATED "C" IN PREDICTION ALGORITHM SET I SENSORS	EPOWER I
Error Source	N1 Limit % Power	T <sub>7</sub> Limit % Power	Wf Limit % Power	Detailed Error List Table No.	Applicable Notes (Appendix I)
Linearization of nonlinear differential equations	0.50	1.25	0.35	иш	-
Nonstandard day	0.10	0.23	0.21		2,3
Use of base engine "C" for degraded engine	0.03	JS*0	0.03	×	4,5
High-power degradation being different from low- power degradation	0.22	0.68	0.22	II,	9
Change in actual power due to uncertainty in WBL and SPE at high power	1.17	4.96	2, 15	и"пх	2
Uncertainty in WBL and SPE at low power	0.37	1.76	0.78	XIX	œ
Sensor errors at low power	2.61	3.60	2.33	хv, хvı, п	G
Control limit and sensor errors at high power (RSS) Total Error	0.52	1.55	$\frac{0.74}{3.38}$	XV, XVI, II	ത

	TABLE V	TABLE V - CONTINUED	ED		
Error Source	N1 Limit % Power	Detailed Error T <sub>7</sub> Limit W <sub>f</sub> Limit List % Power % Power Table No.	W <sub>f</sub> Limit % Power	Detailed Error List Table No.	Applicable Notes (Appendix I)
Standard Sensors:			ı		
Sensor Errors at low power	6.5	7.02	6.46	ху, хуп, п	10
Control limit and sensor errors at high power	0.52	1.55	0.99	ху, хуп, п	10
(RSS) Total Error	7.15	9.03	6.53		

TABLE VI. ERROR N2 = 10 ALGOR	ERROR SUMMARY SHEET - N2 = 100% USING INTERPOLA ALGORITHM SET I SENSORS	HEET - PR TERPOLATI SENSORS	EDICTION	ERROR SUMMARY SHEET - PREDICTION AT 19.7% POWER AT N2 = 100% USING INTERPOLATED "C" IN PREDICTION ALGORITHM SET I SENSORS	R AT
Error Source	N <sub>1</sub> Limit % Power	T7 Limit % Power	Wf Limit % Power	Detailed Error List Table No.	Applicable Notes (Appendix I)
Linearization of nonlinear differential equations	0.50	1.23	0.35	иш	1
Nonstandard day	0.10	0.23	0.21		2,3
Use of base engine "C" for degraded engine	0.03	0.50	0.03	×	4,5
High-power degradation being different from low- power degradation	0.22	0.68	0.22	₩	9
Change in actual power due to uncertainty in WBL and SPE at high power	1.17	4.96	2.15	хип, п	٢
Uncertainty in WBL and SPE at low power	0.36	1.77	0.76	XIV	œ
Sensor errors at low power	4.20	5.06	3,93	ху, хуш, п	6
Control limit and sensor errors at high power	0.52	1.55	0.74	ху, хуш, п	თ
(RSS) Total Error	4.44	7.62	4.63		

TABLE VII. ERROF AT N <sub>2</sub> ALGOF	ERROR SUMMARY SHEET - I AT N2 = 100% USING INTERPA ALGORITHM SET II SENSORS	SHEET - PR INTERPOI SENSORS	EDICTION ATED "C"	ERROR SUMMARY SHEET - PREDICTION AT 35.2% HORSEPOWER AT N <sub>2</sub> = 100% USING INTERPOLATED "C" IN PREDICTION ALGORITHM SET II SENSORS	RSE POWER ON
Error Source	N1 Limit % Power	T <sub>7</sub> Limit % Power	Wf Limit % Power	Detailed Error List Table No.	Applicable Notes (Appendix I)
Linearization of nonlinear differential equations	0.50	1.23	0.35	VIII	1
∂A <sub>N</sub> /A <sub>N</sub> not equal to -∂ηρt /ηρt	0.30	0.54	0.40		11
Nonstandard day	0.10	0.23	0.21		2,3
Use of base engine "C" for degraded engine	0.03	0.50	0.30	×	4,5
High-power degradation being different from low-power de- gradation	0.22	0.68	6.22	ПX	9
Change in actual power due to uncertainty in WBL and SPE at high power	1.17	4.96	2.15	хіні, ш	<b>L</b> -
Uncertainty in WBL and SPE at low power	0,38	1.70	0.78	ΛΙΧ	œ
Sensor Errors at low power	2.34	3.17	2.02	XV, XIX	æ

TABI	TABLE VII. Continued	tinued			
Error Source	Nj Lim % Power	$egin{array}{c} \mathbf{T_7} \ \mathbf{Lim} \ \% \ \mathbf{Power} \end{array}$	<sup>W</sup> f Lim % Power	Detailed Error List Table No.	Applicable Notes (Appendix I)
Control Limit and sensor errors at high power	0.52	1.55	0.74	XV, XIX, III	6
Total Error	2.77	6.52	3.20		
Standard Sensors:					
Sensor errors at low power	5.66	3.65	5.49	XV,XX,III	10
Control limit and sensor errors at high power	0.52	. 55	0.29	ху,хх,ш	10
lotal Erfor	es •c	9.7.4	6 <b>.</b> U?		

(computed at maximum power on a standard day) results in a significant prediction error at nonstandard-day conditions. Table V is an error summary having the same conditions as Table IV except that all "C" are interpolated values. Details about interpolation of "C" are described later. Table VI is an error summary using the same algorithm as in Table V, but predicting maximum power at a lower horsepower (19.7%). Table VII is an error summary when using Set II measurements at low power.

Tables IV through VII summarize the root sum squared (RSS) prediction error for each of the three control limits (N<sub>I</sub>, T<sub>7</sub>, and W<sub>f</sub>). Tables VIII through XX contain a detailed error list. Also, Appendix I provides additional information and the conditions assumed when performing the error analysis. Tables V and VII also contain an error summary when using so-called "standard sensors". In particular, the pressure sensors and shaft horsepower sensors are not the "best available." Errors in the assumption that  $\partial A_N/A_N = -\partial \eta_{pt}/\eta_{pt}$  are a source of power prediction error and are included in the error summary sheet.

The range of ambient conditions where each control limit influences maximum power is defined in Figure 18. This figure shows the combinations of  $P_l$  and  $T_l$  where the  $N_l$  limit,  $T_7$  limit, and  $W_f$  limit determine the maximum power. For example, horsepower is limited by  $W_f$  at high  $P_l$  and limited by  $N_l$  at low  $P_l$  in the temperature range of  $-12^{\circ}F$  to  $59^{\circ}F$  for the base engine. All three limits occur simultaneously at standard-day conditions ( $P_l = 14.7$  psia and  $T_l = 518.7^{\circ}R = 59^{\circ}F$ ). The range of thermodynamic conditions where a  $W_f$  limit,  $T_7$  limit, or  $N_l$  limit can occur is smaller than the complete range of ambient conditions. Hence, the desired changes in the influence of horsepower on coefficients "C" are correspondingly smaller and are listed in Table IX. The effect of nonstandard day on "C" in the range where each limit can be encountered is shown in Table X.

It is noted in Table IV that a dominant error is caused by the use of a constant "C" (computed at standard day) for nonstandard-day conditions. Studies indicate that most of this error can be eliminated by storing two values of each "C" (instead of one) and then interpolating as a function of the existing  $P_1$  and  $T_1$ . The specific conditions where "C" is computed (and stored) and the method of interpolation for each limit are defined below:

Maximum         Base Engine "C"         hp Error at ±5% "C"           Variable         Variation         N₁ Limit         Tγ Limit         Wf Limit         Tγ Limit         Wf Limit           η <sub>c</sub> -3.5        7620         3.3798         .8269         .133         .59         .145           η <sub>t</sub> -2.5        9790         4.3485         1.859         .122         .544         .148           Λ <sub>5</sub> -2.5         1.0044         1.0061         1.0051         .176         .126         .126           A <sub>5</sub> +2.2         -1.939         -1.5644         -6.320         .089         .176         .004           Wa         -2.0         1.2929        0597        1159         .129         .006         .016           Wa         -2.0         1.2929        0597        1159         .129         .006         .016           η <sub>7</sub> -2.5        7620         3.3798         .8269         .135         .87         .145           η <sub>7</sub> -2.5        9790         4.3485         1.1859         .122         .544         .148           η <sub>7</sub> -2.5        9790         4.3485         1.1859		TAI	BLE VIII, DE	TAILED ERR NLINEAR DI	DETAILED ERROR LIST - LINEARIZATION OF NONLINEAR DIFFERENTIAL EQUATIONS	VEARIZATIO EQUATIONS	N OF	
Variation N <sub>1</sub> Limit T <sub>7</sub> Limit W <sub>f</sub> Limit R <sub>D</sub> Power R <sub>D</sub> Power  -3.57620 3.3798 .8269 .133 .59  -2.5 1.0044 1.0061 1.0051 .176 .126  +2.2 .7930 -1.56446320 .089 .176  +2.2 .7930 -1.56446320 .089 .176  -2.0 1.292905971159 .129 .267  -3.57620 3.3798 .8269 .135 .59  -3.57620 3.3798 .8269 .132 .54  -2.5 1.0044 1.0061 1.0051 .176 .126  +2.2 .7930 4.3485 1.1859 .122 .54  -2.5 1.0044 1.0061 1.0051 .176 .126  +2.2 .7930 2.4311034 .186 .267  -2.5 1.0044 1.0061 1.0051 .176 .172  +2.2 .1637 2.4311034 .186 .267  -2.0 1.292905971159 .129  RSS Error 3.5 .87  RSS Error 3.5 .87  RSS Error 3.5 .87  RSS Error 3.5 .87		Maximum	Be	ase Engine "C	H.	I dq	rror at ±5%	'C''
## Power ##	Variable	Variation		T <sub>7</sub> Limit		N <sub>l</sub> Limit	T <sub>7</sub> Limit	Wf Limit
-3.57620 3.3798 .8269 .133 .59 -2.5 1.0044 1.0061 1.0051 .176 .126 +2.2 .7930 -1.56446320 .089 .176 +2.2 -1.6937 2.4311034 .186 .267 -2.0 1.292905971159 .129 .006 -2.59790 4.3485 1.1859 .122 .594 -2.59790 4.3485 1.1859 .122 .594 -2.5 1.0044 1.0061 1.0051 .176 .126 +2.2 .7930 -1.56446320 .087 .172 +2.2 .7930 -1.56446320 .087 .172 -2.5 1.024 1.0061 1.0051 .176 .126 +2.2 .7930 -1.56446320 .087 .172 -2.5 1.0944 1.0061 1.0051 .176 .126 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0944 1.0061 1.0051 .176 .128 -2.5 1.0044 1.0061 1.0051 .176 .128 -2.5 1.0044 1.0061 1.0051 .176 .128 -2.5 1.0044 1.0061 1.0051 .176 .128 -2.5 1.0044 1.0061 1.0051 .176 .128 -2.5 1.0044 1.0061 1.0051 .176 .128 -2.5 1.0044 1.0061 1.0051 .176 .128 -2.5 1.0044 1.0061 1.0051 .176 .186 .267 -2.5 1.0044 1.0061 1.0051 .186 .267 -2.5 1.0044 1.0061 1.0051 .186 .267 -2.5 1.0044 1.0061 1.0051 .186 .267 -2.5 1.0044 1.0061 1.0051 .186 .267 -2.5 1.0046 1.0061 1.0051 .186 .267 -2.5 1.0046 1.0061 1.0051 .186 .267 -2.5 1.0066 1.288 1.0066 1.		2/2	,		<b>'</b>	% Power	% Power	% Power
-2.59790	, u	-3,5	-, 7620	3,3798	. 8269	.133	. 59	.145
-2.5 1.0044 1.0061 1.0051 1.76 1.126 +2.2 7930 -1.56446320 0.089 1.76 +2.2 1.6937 2.4311034 1.186 1.267 -2.0 1.292905971159 1.29 -3.57620 3.3798 8.8269 1.32 5.94 -2.59790 4.3485 1.1859 1.122 5.54 -2.5 1.0044 1.0061 1.0051 1.76 1.126 +2.2 7930 -1.56446320 0.087 1.172 +2.2 1.6937 2.4311034 1.186 1.267 -2.0 1.292905971159 1.29 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28 -2.0 1.292905971159 1.28	, <u>u</u>	-2.5	9790	4,3485	1,1859	.122	. 544	.148
+2, 2 . 7930	7 pt	-2.5	1,0044	1,0061	1,0051	.176	.126	.126
+2, 2 -1,6937 2,4311034 .186 .267 -2, 0 1,292905971159 .129 .006  -3, 57620 3,3798 .8269 .132 .59 -2, 59790 4,3485 1.1859 .122 .544 -2, 5 1,0044 1,0061 1,0051 .176 .126 +2, 2 .7930 -1.56446320 .087 .172 +2, 2 .16937 2,4311034 .186 .267 -2, 0 1,29290597 .1159 .129 .006  RSS Error 35 .87  error in "C" and ±5% ard ±5% ard ±5% ard ±5% "B"	A5	+2.2	.7930	-1.5644	6320	680.	.176	070.
-2.0 1.292905971159 .129 .006  RSS Error 35 .87  -3.57620 3.3798 .8269 .132 .59  -2.59790 4.3485 1.1859 .122 .544  -2.5 1.0044 1.0061 1.0051 .176 .126  +2.2 .7930 -1.56446320 .087 .172  +2.2 -1.6937 2.4311034 .186 .267  -2.0 1.292905971159 .129 .006  RSS Error 35 .87  RSS Error 35 .87  error in "C"  and ±5%  error in "B".	AN	+2.2	-1,6937	2,4311	034	.186	. 267	.004
RSS Error .35 .87  -3.57620 3.3798 .8269 .132 .59  -2.59790 4.3485 1.1859 .122 .544  -2.5 1.0044 1.0061 1.0051 .176 .126  +2.2 .7930 -1.56446320 .087 .172  +2.2 .1.6937 2.4311034 .186 .267  -2.0 1.292905971159 .129  RSS Error .35 .87  RSS Error .35 .87  and ±5% error in "C"  and ±5% error in "B".	Wa	-2.0	1,2929		-,1159	129	900	.016
hp Error at ±5% "B"  -3.5 7620 3.3798 .8269 .132 .59  -2.5 9790 4.3485 1.1859 .122 .544  -2.5 1.0044 1.0061 1.0051 .176 .126  +2.2 .7930 -1.56446320 .087 .172  +2.2 -1.6937 2.4311034 .186 .267  -2.0 1.292905971159 .129  RSS Error .35 .87  RSS Error .50 1.28  and ±5%  error in "C"  and ±5%  error in "B".					RSS Error	.35	. 87	. 25
-3.57620 3.3798 .8269 .132 .59 -2.59790 4.3485 1.1859 .122 .544 -2.5 1.0044 1.0061 1.0051 .176 .126 +2.2 .7930 -1.56446320 .087 .172 +2.2 -1.6937 2.4311034 .186 .267 -2.0 1.292905971159 .129 .006  RSS Error .35 .87  RSS Error and ±5% error in "C" and ±5% error in "C" and ±5%						4	rmor at +5%	<u>ئ</u> ت
-2.59790 4.3485 1.1859 .122 .544 -2.5 1.0044 1.0061 1.0051 .176 .126 +2.2 .7930 -1.56446320 .087 .172 +2.2 -1.6937 2.4311034 .186 .267 -2.0 1.292905971159 .129 .006 RSS Error .35 .87  RSS Error and ±5% error in "C" and ±5% error in "B".	<b>1</b>	-3.5	7620	3,3798	. 8269		. 59	
-2.5 1.0044 1.0061 1.0051 .176 .126 +2.2 .7930 -1.56446320 .087 .172 +2.2 -1.6937 2.4311034 .186 .267 -2.0 1.292905971159 .129 RSS Error .35 .87  RSS Error and ±5% error in "C" and ±5% error in "B".	n,	-2.5	0646 -	4,3485	1,1859	.122	.544	.148
+2.27930 -1.56446320 .087 .172 +2.2 -1.6937 2.4311034 .186 .267 -2.0 1.292905971159 .129 .006 RSS Error .35 .87 RSS Error and ±5% error in "C" and ±5% error in "B".	npt	-2.5	1,0044	1,0061	1,0051	.176	.126	.126
+2.2 -1.6937 2.4311034 .186 .267 -2.0 1.292905971159 .129 .006  RSS Error .35 .87  RSS Error and ±5% error in "C" and ±5% error in "B".	A5	+2.2	. 7930	-1.5644	-, 6320	.087	.172	.070
-2.0 1.292905971159 .129 .006  RSS Error .35 .87  RSS Error .50 1.28  due to ±5% error in "C" and ±5% error in "B".	AN	+2.2	-1.6937	2,4311	034	.186	. 267	.004
RSS Error .35 .87  RSS Error .50 1.28  due to ±5%  error in "C"  and ±5%  error in "B".	Wa	-2.0	1.2929	0597	-,1159	129	900	.016
.50 1.28					RSS Error	.35	. 87	. 25
.50 1.28								
due to ±5%  error in "C"  and ±5%  error in "B".					RSS Error	.50	1.28	.35
error in "C" and ±5% error in "B".					due to ±5%			
and ±5% error in "B".					error in "C	•		
error in "b".					and ±5%			
					error in "B"	•		

TABLE IX. DETAILED ERROR LIST - EFFECT OF NONSTANDARD DAY ON THE VALUE OF "C" BASE ENGINE N<sub>1</sub> Limit  $T_1 = 59^{\circ} F$  $T_1 = -12^{\circ}F$  $N_{1C} = 24,700$  $N_{1C} = 26,600$ Δ"C" Variable "c " " C"  $\eta_{\mathrm{c}}$ - .7620 + .0330 - .7950  $\eta_{\mathrm{t}}$ -.9790-1.0583+ .0793  $\eta_{
m pt}$ - .0017 1.0044 1.0061  $A_{N}$ - .0560 -1.6937-1.6377 $\mathbf{w_a}$ 1.2929 1.3710 - .0781  $A_5$ .7930 .7579 + .0351 T7 Limit  $T_1 = 59^{\circ} F$  $T_1 = 120^{\circ} F$  $N_{1C} = 24,700$  $N_{1C} = 22,900$ "C " " C "  $\Delta^{"}C"$ Variable  $\eta_{c}$ 4.3462 - .9664 3.3798  $\eta_{t}$ 4.3485 5.5683 -1.2198**7**pt 1.0061 + .0009 1.0052 - .7338 2,4311 3.1649  $A_N$  $w_a$ - .3484 + .2887 - .0597 **-2.0589**. + .4945  $A_5$ -1.5644

		TABLI	TABLE IX - Continued		
		W <sub>f</sub> Limit			
	$T_1 = 59^{\circ}F$ $N_1C = 24,700$ $P_1 = 14.7 psia$	$T_1 = -60^{\circ}F$ $N_1C = 26,900$ $P_1 = 11$	$T_1 = 72^{\circ}F$ $N_1C = 24,340$ $P_1 = 16$		
Variable	° °	: O :	: O :	" C " (Col. 1 - Col. 2)	"C" (Col. 1 - Col. 3)
P1/P1Ref	-, 8409	6377	- 8835	-, 2032	.0426
7°C	.8269 1.1859	.6351	. 8758 1, 2533	.1918	-, 0489
7 PT	:.0051	1.0071	1,0047	-,0020	+000+
A <sub>N</sub>	-, 0340 -, 0281	1105 +.0727	.0058	+, 0765 -, 1008	0398 +. 0198
A 5	-, 4873	-, 3507	-, 5209	1366	+, 0336

FAB	FABLE X. DETAILED CAUSED BY	ERROR LIST -	DETAILED ERROR LIST - PREDICTION ERROR AT NONSTANDARD DAY CAUSED BY USING CONSTANT "C" ("C" COMPUTED AT STANDARD DAY	OR AT NONST PUTED AT ST	ANDARD DAY
	PREDICT	PREDICTION AT COLDEST DAY	ST DAY		
Variable	Maximum $\Delta$ From Nominal	Speed Limit $(T_1 = -12^{\circ}F)$ $\Delta$ "C" hp Error	$(T_1 = -12^{\circ}F)$ hp Error	Fuel Limit $\Delta$ C."	Fuel Limit (T <sub>1</sub> = $-60^{\circ}$ E) $\Delta$ "C" hp Error
P1/P1REF	17%			2032	3, 454
″c	3.5%	+. 0330	+. 116	.1918	.671
/ <sub>t</sub>	2.5%	+, 0793	.198	. 2813	. 703
7 pt	2.5%	0017	.004	0020	. 005
AN	2.2%	0560	. 123	+.0765	. 168
Wa	2.0%	0731	.156	1008	. 202
A5	2.2%	+, 0351	. 077	1366	.301
			. 313% RSS		3.61% RSS
			Error		Error
	PREDICT	PREDICTION AT HOTTEST DAY	ST DAY		
	Maximum $\Delta$	Temp. Limit	Temp. Limit (T <sub>1</sub> = $120^{\circ}$ F)	Fuel Limit (T1	$(T_1 = 72^{\circ}F)$
Variable	From Nominal	Δ "C"	hp Error	Δ"С"	hp Error
P <sub>1</sub> /P1REF	17%			. 0426	. 724
″c	3.5%	9664	3.382	0489	.171
"t	2.5%	-1.2198	3.049	0674	. 169
7 pt	2.5%	+* 0000	. 002	£003.	. 001
AN	2.2%	7338	1.614	0398	880.
Wa	2.0%	.2887	.577	.0198	. 040
A5	2.2%	. 4945	1.088	. 0336	. 074
			4.98% RSS		. 772% RSS
			Error		Ermr
				1	

		TABLE XI.	- 1	AILED ERRC	DETAILED ERROR LIST - USE OF BASE ENGINE "C" FOR DEGRADED ENGINE	OF BASE	
Limit	Variable	Defined Degradation	Base Engine "C"	Degraded Engine "C"	Average Value of	Error in "C"	Horsepower Error % Power
ź	ηc	-1.	7620	-, 7555	-, 7583	- 0033	0033
4	7 t	-1:	9790	9564	9677	0113	. 0113
	7 pt	-1:	1.0044	1.0038	1.0041	. 0003	. 0003
	$A_5$	+1.	. 7930	. 8019	. 7975	0045	.0045
	$A_N$	+1.	-1.6937	-1.7164	-1.7051	.0114	.0114
						RSS ERROR	OR 0.017
$ T_2 $	ηc	-1.	3.3798	3.6521	3.5160	1362	.1362
	7 t	-1.	4.3485	4.6728	4.5106	1621	.1621
	7 pt	-1 <b>-</b>	1,0061	1.0055	1,0058	. 0003	. 0003
	A5	+1.	-1.5644	-1.7113	-1.6379	.0735	.0735
	$A_N$	+1.	2.4311	2.6632	2.5471	1160	.1160
						RSS ERROR	
Wf	ηc	-1.	3.269	. 8454	.8361	0092	. 0092
	$\frac{\eta}{t}$	-1.	1.1859	1.2045	1.1952	0093	. 0093
_	7 pt	-1.	1,0051	1.0047	1.0049	. 0002	. 0002
	$A_5$	<del>,</del> 1.	4873	5007	4940	. 0067	. 0067
	AN	+1.	0340	0323	0331	0009	6000
						RSS ERROR	OR 0.015
NOTES:	.;						
	Consider horsepowe	Consider a "Severe Degradation horsepower error is as follows:	radation" follows:	to be twice th	ıe above "Defin	a "Severe Degradation" to be twice the above "Defined Degradation". or error is as follows:	. The estimated
	$M_{\rm f}$	Limit Error = $0.034\%$ Limit Error = $0.30\%$	0.034%	T <sub>7</sub> Limit	$T_7$ Limit Error = 0.50%		
	1						

TABLE XII. DETAILED ERROR LIST - HIGH POWER DEGRADATION BEING DIFFERENT FROM LOW POWER DEGRADATION

Limit	Variable	Base Engine ''C''	Estimate Difference Between Low and High Power Degradation	Horsepower Error
N <sub>1</sub>			(%of Point)	
``1	η <sub>c</sub>	7620	± <b>0.15</b>	.1143
	$\eta_{\mathbf{t}}$	9790	± 0.10	.0979
	<sup>7</sup> pt	1.0044	± 0.10	.1004
	$W_{\mathbf{a}}$	1.2929	± <b>0.10</b>	.1293
	A <sub>5</sub>	.7930	0.	0.
	An	-1.6937	0.	0
			RSS Error	0.22%
Т7	7 <b>c</b>	3.3798	± <b>0.1</b> 5	.5070
•	7 t	4.3485	± <b>0.1</b> 5	.4349
	"pt	1.0061	± 0.10	.1006
	W <sub>a</sub>	0597	+ 0.10	.0060
	A <sub>5</sub>	-1,5644	0.	0.
	A <sub>n</sub>	2.4311	0.	0.
			RSS Error	0.68%
$w_{\mathbf{f}}$	η <sub>c</sub>	. 8269	± 0.15	.1240
•	$\eta_{\mathbf{t}}$	1.185 <b>9</b>	± 0.10	.1186
	<sup>7</sup> pt	1.0051	± 0.10	.1005
	$\mathbf{w_a}$	0281	± 0.10	.0028
	A <sub>5</sub>	6320	0.	0.
	An	034	0.	0.
			RSS Error	0.20%

DETAILED ERROR LIST - UNCERTAINTY IN COMPRESSOR WBL AND SPE AT MAXIMUM POWER WHEN PREDICTING AT LOW POWER	ver	W <sub>f</sub> Limit	% Power	2,145	204	2,15%
SSOR WBI	Actual Horsepower	Variation T <sub>7</sub> Limit	% Power % Power % Power	1.17 4.935	.07	4.96%
A COMPRE LOW POWI	Actua	N <sub>l</sub> Limit	% Power	1.17	. 07	1.17%
TAINTY D		Variation No Limit To Limit We Limit No Limit To Limit We Limit	,,C,,	1.43	68	or
r – uncer en predic		T7 Limit	,,C,,	3, 29	-1.57	RSS Error
RROR LIST		N <sub>1</sub> Limit	,,C,,	.78	. 22	
			Uncertainty	5% of max. Wa	±.30 hp/pps of Wa	
тавсе хп.			e		±, 3	
TABL			Variable	WBL	SPE	

	TABI	TABLE XIV. DE'	TAILED E	DETAILED ERROR LIST - UNCERTAINTY IN COMPRESSOR W <sub>BL</sub> AND SPE AT LOW POWER	T - UNCEF D SPE AT	TAINTY I	n Ær	
Sensors & Power per Table	Variable	Sensitivity hp Prediction Error Variable Uncertainty $N_1$ Limit $T_7$ Limit $W_{f f}$ Limit $W_{f f}$ Limit	N <sub>I</sub> Limit	Sensitivity T <sub>7</sub> Limit	W <sub>f</sub> Limit	hp Pre N <sub>l</sub> Limit	hp Prediction Error Limit T <sub>7</sub> Limit W <sub>f</sub>	ror W <sub>f</sub> Limit
I or II	WBL SPE	±.5% ±.3 hp/pps	.71	3. 24 2. 26 RSS Error	1,46 ,95	$\frac{35}{37\%}$	1.62 .68 1.76%	. 73 . 28 . 78%
E	$^{ m WBL}_{ m SPE}$	±.5% ±,3 hp/pps	. 66	3.26 2.60 RSS Error	1.37 1.07 or	$\begin{array}{c} \textbf{.33} \\ \textbf{.144} \\ \textbf{.36} \% \end{array}$	$\frac{1.63}{1.77\%}$	.685 .32 .76%
2	$^{ m WBL}_{ m SPE}$	±, 5% ±, 3 hp/pps	.72	3, 16 2, 13 RSS Error	1,43 ,969 or	$\frac{36}{12}$	$\frac{1.58}{.64}$	. 29 . 78%

	TABLE XV.	STEADY-ST	STEADY-STATE SENSOR AND CONTROL ACCURACY ESTIMATES	NTROL ACCURACY I	ESTIMATES
				Best	Standard
				Available	Sensors
	Range	Range		Error Including	Error Including
Sensor	Min.	Max.	Sensor Type	Sensor Conversion	Sensor Conversion
Ż	20K mm	95K mm	Tachometer	+2 5 mm	4.0 5 mm
I Z	13K rpm	mdr Moc	Tochomotor	10. tpm	+ 10 thur
N .	100°TO	Tooo	Datime Designation	±2.0 Ipm	±2.5 rpm
1 T	400 K	1050 K	Charles Resistance	#4. c R	±2.2 K
 ಬ	670 K	10001	Curomel -Alumel	±6, 3 K	±6.3 K
			(Thermocouple)		
T7	$1100^{\circ}\mathrm{R}$	$1800^{\circ}\mathrm{R}$	Chromel-Alumel	±6,3°R	±6.3°R
T9	$1000^{\circ}\mathrm{R}$	$1500^{\circ}\mathrm{R}$	Chromel-Alumel	±6,3°R	±6.3°R
P <sub>1</sub>	ll psia	16 psia	Strain Gage		±, 19 psi
P3	50 psia	110 psia	Strain Gage		±1,05 psi
P7	20 psia	30 psia	Strain Gage		±.36 psi
$P_1$	11 psia	16 psia	Vibrating Cylinder	±.02 psi	
P <sub>3</sub>	50 psia	110 psia	Vibrating Cylinder	±, 10 psi	
P7	20 psia	30 psia	Vibrating Cylinder	±.05 psi	
Wf	350 pph	800 pph		±2 pph at 400 pph	±2 pph at 400 pph
				±4 pph at 800 pph	±4 pph at 800 pph
SHP	300  hp	1500 hp		±10 hp	±30 hp
	300 hp	200 hp			
Control					
N <sub>1</sub> SET	= 2.2 (N)	NISET/ T1)		21 rpm	21 rpm
T7 SET				0	. 0
Wf SET	= .19 (				±2.55 pph
	= .02 ( W	Wf SET/ P1)		±.27 pph	

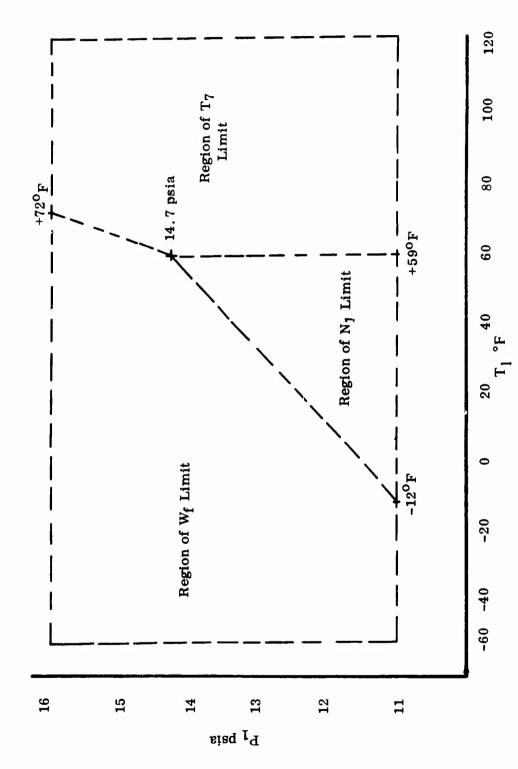
TABLE XVI.		DETAILED ERROR LIST - SENSOR ERRORS AT 35, 2% HORSEPOWER AND AT MAXIMUM HORSEPOWER USING SENSORS AT LOW POWER AS TABULATED BEST AVAILABLE SENSORS	R LIST - S POWER U E SENSORS	ENSOR ER SING SENS	RORS AT	35. 2% HO	SENSOR ERRORS AT 35, 2% HORSEPOWER AND AT USING SENSORS AT LOW POWER AS TABULATED AS	R AND AT ULATED
		Error in		Sensitivity		hp Pre	hp Prediction Error	ror
Sensor	Error	% of Point	N <sub>l</sub> Limit	T <sub>7</sub> Limit	W <sub>f</sub> Limit	N <sub>l</sub> Limit	T <sub>7</sub> Limit	Wf Limit
Low Power:								
N I	±2, 5 rpm	±.012	-7.45	4.02	-2,88	.0894	.0482	.0346
T3	±6.3°R	±.71	14	10	19	. 0994	. 071	.1349
Wf	±2 pph	±.46	56	+ .01	-1.46	. 2714	.0046	.6716
SHP	±10 hp	±2.08	1.0	1.0	1.0	2,08	2,08	2,08
$\Gamma_7$	±6.3 R	±.43	20	4.4	03	.0860	1,872	.0129
$P_7$	$\pm$ 05 psi	±, 19	39	71	52	.074	.1349	8860.
N <sub>2</sub>	±2.5 rpm	±,013	1,03	1,03	1.03	.0134	.0134	.0134
$T_1$	±2, 2'R	±.42	3.68	5,36	1.79	1,545	2, 251	. 7518
$_{ m l}$	±. 02 psi	±, 135	45	09 -	.84	. 0608	.081	.1134
$P_3$	±. 10 psi	±, 143	. 21	.14	80.	. 0300	.0200	. 0114
				RSS Error	or	2,61	3,60	2, 33
Maximum Power:	:Ic							
Z Z	±2.5 rpm	± <b>.</b> 01	-3, 71	0	0	. 04	0	0
J.M	±4 pph	<b>±.</b> 50	• 56	0	1,47	. 28	0	. 735
T7	±6.3°R	±.36	0	-4.31	0	0	1,552	0
${\mathtt T}_{\mathtt J}$	±2.2°R	±.42	92	0	0	.319	0	0
$P_1$	$\pm$ . 02 psi	±, 143	37	0	0	.0529	0	0
N <sub>1</sub> SET	±21 rpm	<b>∓°</b> 08	3,71	0	0	. 2968	0	0
T <sub>7</sub> SET	0	0	0	4.31	0	0	0	0
W <sub>f</sub> SET	±. 27 pph	±. 034	0	0	-1.47	0	0	.0500
				RSS Error	or	. 52%	1.55%	.74%

TABLE XVII.		DETAILED ERROR LIST - SENSOR ERRORS AT 35, 2% HORSEPOWER MAXIMUM POWER USING SENSORS AT LOW POWER AS TABULATED STANDARD SENSOR ACCURACY	R LIST - S R USING SI OR ACCUR	- SENSOR ERRORS NT 35, 2% HORSEPOWER AND AT 3 SENSORS AT LOW POWER AS TABULATED URACY	RORS AT	35. 2% HOR WER AS TA	SE POWEF ABULA TE	t AND AT
		Error in		Sensitivity		hp Pre	hp Prediction Error	ror
Sensor	Error	% of Point	N <sub>1</sub> Limit	T <sub>7</sub> Limit	Wf Limit	N <sub>l</sub> Limit	T <sub>7</sub> Limit	Wf Limit
Low Power:								
N		±, 012	-7.45	4.02	-2.88	.0894	.0482	.0346
T3	±6.3°R	±.71	14	10	19	.0994	. 071	.1349
Wf	±2 pph	<b>±.46</b>	59	+ .01	-1.46	. 2714	.0046	.6716
SHP	<b>430 hp</b>	<b>±6.</b> 25	1.00	1.00	1.00	6, 25	6, 25	6.25
$\Gamma_7$	±6.3°R	±.43	20	4.4	03	0980	1.872	. 0129
$P_7$	±.36 psi	±1,36	39	71	52	. 5304	. 9656	. 7072
N <sub>2</sub>	±2.5 rpm	±, 013	1.03	1,03	1,03	.0134	.0134	.0134
T	±2.2°R	±.42	3.68	5, 36	1.79	1.545	2, 251	.7518
Pl	±. 19 psi	±1,29	45	09	. 84	. 5405	.774	1.0836
P3	±1.05 psi	±1.50	. 21	.14	80.	.315	. 21	.12
				RSS Error	or	6.50%	7.02%	6.46%
Maximum Power:	ü							
N	±2.5 rpm	₹, 01	-3, 71	0	0	.04	0	0
JM	±4 pph	± <b>.</b> 50	• 56	0	1.47	. 28	0	. 735
T.I.	±6.3°R	±,36	0	4.31	0	0	1.552	0
$\Gamma_{l}$	±2.2°R	±.42	92	0	0	.319	0	0
Pl	±, 19 psi	±1.29	0	0	37	0	0	.4773
N <sub>1</sub> SET	±21 rpm	<b>∓°</b> 08	3,71	0	0	. 2968	0	0
T7 SET	0	0	0	4.31	0	0	0	0
Wf SET	±2.55 pph	±,32	0	0	-1.47	0	ပ	4704
				RSS Error	or	. 52%	1.55%	%66°

TABLE XVIII.		ERROR LIST - SENWER AS TABULATED AILABLE SENSORS	T - SENSO LATED ISORS	R ERROR	S AT 19, 7%	6 HORSE PO	WER USIN	DETAILED ERROR LIST - SENSOR ERRORS AT 19.7% HORSEPOWER USING SENSORS AT LOW POWER AS TABULATED BEST AVAILABLE SENSORS
Sensor	Error	Error in $\%$ of Point	N <sub>l</sub> Limit	Sensitivity T <sub>7</sub> Limit	W <sub>f</sub> Limit	Error in Sensitivity hp Prediction Error $\%$ of Point N1 Limit T7 Limit Wf Limit N1 Limit T7 Limit Wf Limit	hp Prediction Error Limit T $_7$ Limit W $_{ m f}$	ror W <sub>f</sub> Limit
Low Power:								<u>.</u>
Z	±2.5 rpm	±,012	10.08	7,32	5, 83	.1210	.0878	.0700
Т3	±6.3°R	±.74	. 24	.05	.10	.1776	.0370	.0740
Wf	±2 pph	÷.56	09.	.03	1.46	.336	.0168	.817
SHP	áy 01~	±3.62	1.00	1.00	1.00	3,62	3,62	3,62
T7	±6.3°R	<b>±.46</b>	.30	4.52	.07	.138	2,079	.032
P7	±. 05 psi	±, 21	. 59	1.23	. 85	. 1239	. 2583	.1785
N <sub>2</sub>	±2.5 rpm	±, 013	2,10	2,10	2,10	. 0273	.0273	.0273
$T_1$	±2.2°R	±.42	4.97	6.77	3.02	2,087	2,843	1,268
P <sub>l</sub>	±. 02 psi	±, 135	.38	90.	1.12	.0513	.0081	.1512
P3	±, 10 psi	±.17	.21	60.	90.	.02295	. 0122	.0081
				RSS Error	or	4.20	5.06	3,93

	BE		BEST AVAILABLE SENSORS	SORS				
Sensor	Error	Error in $\%$ of Point	N <sub>1</sub> Limit	Sensitivity T <sub>7</sub> Limit	Wf Limit	hp Pre N <sub>l</sub> Limit	hp Prediction Error N <sub>1</sub> Limit T <sub>7</sub> Limit W <sub>f</sub>	ror W <sub>f</sub> Limit
Low Power:						-		
Z	±2,5 rpm	±. 012	-7.08	-3.14	-2.44	.0849	.0377	.0293
$T_{5}$	$\pm 6$ , $3$ °R	±. 71	12	01	90	.0852	.0071	.0426
J <sub>W</sub>	±2 pph	±.46	09	+ .54	-1.49	. 276	. 2484	. 685
SHP	<b>du</b> 0 1∓	±2,08	. 86	. 23	98.	1.79	.48	1.79
$T_7$	$\pm 5.3^{\circ} \mathrm{R}$	<b>±.</b> 50	90 -	-4.52	.17	. 03	2,26	60.
$_{53}^{\rm N}$	±2,5 rpm	±, 013	88.	. 73	.84	. 0114	.0095	.0109
$T_1$	±2, 2'R	±.42	3,52	5.15	1.46	1,478	2,16	.613
$P_1$	±.02 psi	±, 135	68	91	+ .51	. 0918	.1229	6890.
P3	±.10 psi	±.143	. 22	.10	. 08	. 0315	.0143	.0114
				RSS Error	or	2,34	3,17	2.02
Maximum Power:	r:							
$\sum_{i}^{N}$	±2, 5 rpm	±, 01	-3,71	0	0	. 04	0	0
Wf	±4 pph	±. 50	• 56	0	1.47	. 28	0	.735
$T_7$	±6.3°R	<b>±.</b> 36	0	-4.31	0	0	1.552	0
$T_1$	±2, 2°R	±.42	92	0	0	.319	0	0
$P_1$	±. 02 psi	±, 143	0	0	37	0	0	.0529
$N_{ m l}$ SET	±21 rpm	₹ 08	3,71	0	0	. 2968	0	0
$T_7$ SET	0	0	4.31	0	0	0	0	0
Wf SET	±. 27 pph	±, 034	0	0	-1.47	0	0	.50
				Dec Bune	3	0		

TABLE XX. DETAILED ERROR LIST - SENSOR ERRORS AT 35, 2% HORSEPOWER AND AT MAXIMUM HORSEPOWER USING SENSORS AT LOW POWER AS TABULATED STANDARD SENSOR ACCURACY	Error in Sensitivity hp Prediction Error Error $\%$ of Point N Limit T7 Limit Wf Limit N Limit T7 Limit Wf Limit	#2.5 rpm  #. 012	1     ±.01     -3.71     0     0     .04     0       ±.50     .56     0     1.47     .28     0       .96     0     1.47     .28     0	76 0 .76 0	#21 rpm #.08 3.71 0 0 .2968 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
TABLE X	Sensor	ver:			N <sub>1</sub> SET T <sub>7</sub> SET W <sub>f</sub> SET



Ambient Conditions Where Each Control Limit Determines Maximum Power of Base Engine. Figure 18.

Control Limit	Corrected Speed at Which "C" Is Computed (rpm)	Variable Used in the Linear Interpolation	Resulting Max.  Error in "C"
$N_1$	24, 700 and 26, 600	$\sqrt{{ t T}_1}$ or $\sqrt{ heta_1}$	Less than ±.01
Т7	22, 900 and 24, 700	${ t T}_1$ or ${ heta}_1$	Less than ±.02
$w_{\mathbf{f}}$	24, 340 and 26, 900	$P_l\sqrt{T_l}$ or $\pmb{\delta}_l\sqrt{\pmb{ heta}_l}$	Less than ±.01

A particularly large prediction error can result from uncertainty in compressor bleed air (WBL). Variations in bleed air at maximum power cause variations in actual maximum power, causing a prediction error. Tables XIII and XIV contain a detailed list of errors caused by  $W_{BL}$  and compressor shaft power extraction (SPE).

Sensor characteristics are listed in Table XV. The sensor range and sensor type assumed in this study are identified in addition to the sensor error. Tables XVI through XX contain detailed lists of power prediction errors resulting from sensor errors.

Review of the error summary sheets (Tables V, VI, and VII) indicates that the total RSS error in power prediction is in excess of the  $\pm 1\%$  goal. However, several improvements in power prediction accuracy are feasible as noted below and summarized in Table XXI:

- 1. The error caused by linearization of nonlinear differential equations can be eliminated by a more sophisticated analytical effort to derive "B" and "C" coefficients to be stored in the power prediction computer.
- 2. The change in actual power can be reduced due to uncertainty in compressor bleed air  $(W_{BL})$  or compressor shaft power extraction (SPE) by providing more nearly constant demands on both  $W_{BL}$  and SPE. Assume that all uncertainties in  $W_{BL}$  and SPE are eliminated.
- 3. The WBL and SPE are either controlled or measured in such a manner at the low-power prediction condition that there is no uncertainty in WBL or SPE.

Table XXI is an error summary sheet for these improved conditions. Note that sensor errors are now the dominant error. A review of the sensor detailed error list (Table XVI) indicates that the dominant errors are caused by  $T_1$ ,  $T_7$ , and

TABLE XXI. ERROR SUMMARY SHEET - PREDICTION AT 35.2% HORSE-POWER AT N<sub>2</sub> = 100% WITH FEASIBLE IMPROVEMENTS IN POWER PREDICTION USING CURRENT "BEST AVAILABLE" SENSORS SET I LOW-POWER SENSORS

Error Source	N <sub>1</sub> Limit % Power	T7 Limit % Power	•
onstandard day	.10	. 23	. 21
se of base engine "C" for			
degraded engine	.03	<b>. 5</b> 0	. 02
igh-power degradation being different from low-power			
degradation	. 22	.68	. 22
Change in actual power due to uncertainty in WBL and SPE			
at high power	0	0	0
ncertainty in WBL and SPE			
at low power	0	0	0
ensor errors at low power	2.61	3.60	2.33
ontrol limit and sensor			
errors at high power	$\frac{.52}{2.67}$	$\frac{1.55}{4.02}$	$\frac{.74}{2.46}$
TOTAL RSS ERROR	2.67	4.02	2.46
rrors in T <sub>1</sub> , T <sub>7</sub> and SHP sensors at low and high			
power	2,63	3,92	2.23
ll prediction errors excluding errors in T <sub>1</sub> , T <sub>7</sub> and SHP			
sensors	49	. 89	1.02
TOTAL RSS ERROR	$\frac{.49}{2.67}$	$\frac{.89}{4.02}$	$\frac{1.02}{2.46}$

SHP sensor errors. The prediction errors caused by these three sensors are listed separately, and all other prediction errors are lumped together (also tabulated in Table XXI). The prediction error caused by ergine degradation, ambient conditions, control errors, and sensor errors is about  $\pm 1\%$  or less when the  $T_1$ ,  $T_7$  and SHP errors are excluded. The power prediction errors caused by  $T_1$ ,  $T_7$  and SHP sensor errors are much more significant than the combination of all other errors.

In summary, the dominant errors in power prediction are caused by the errors in currently available  $T_1$ ,  $T_7$ , and SHP sensors and uncertainty in  $W_{\rm BL}$  and SPE. If it is possible to eliminate the uncertainty in  $W_{\rm BL}$  and SPE by some operational procedure, then the dominant errors in power prediction are caused by sensing  $T_1$ ,  $T_7$ , and SPE.

A sensitivity and error analysis was made for Set III sensors to try to reduce the prediction error due to sensor errors at low power. This set is identical to the original set except that the shaft horsepower measurement is replaced by T9. The sensitivity table for this set of sensors is shown in Table XXII, Table XXIII is a detailed list of the errors due to low-power sensor inaccuracies and uncertainty in WBL and SPE at low power. Table XXIV summarizes the error for the power prediction system using this set of low-power sensors. Instead of reducing the power prediction error, this set of sensors results in a significant increase in the error due to low-power sensor inaccuracy. For this reason, this set of sensors is abandoned as a possible candidate for the final power prediction model.

Set IV, which uses one less sensor than any set previously considered, is identical to Set I but, instead of P7, assumes the relationship  $\partial A_N/A_N = -\partial \eta_{pt}/\eta_{pt}$ . This set is also identical to Set II, with T7 replacing the T9 measurement at low power. Table XXV is the sensitivity table for this set of sensors. Comparing Table XXV with Table III, it is seen that the sensitivities are not significantly different.

#### EFFECT OF POWER CONDITION ON MPA PREDICTION

Tables XXVI and XXVII give a detailed listing of the errors due to Set IV sensor inaccuracies at 35.2% and 19.7% power. Generally speaking, as power predictions are made at low power levels, the error in predicted power increases. The increase in error is due primarily to an increase in the sensitivity to sensor errors and also to an increase in the percentage of point sensor errors. Table XXVIII is a summary of all the errors for Set IV sensors when predicting at 35.2% power. Comparing Table XXVIII with Table VII, it is seen that the total error for both sets of sensors is essentially the same.

TABLE XXII.	1	SENSITIVITY STUDY - EFFECT OF SENSOR ERROR AT AT HIGH POWER ON POWER REMAINING ACCURACY	STUDY - EF	- EFFECT OF SENSOR ERROR AT LOW POWER AND POWER REMAINING ACCURACY SET II SENSORS	SOR ERROR	AT LOW POWER A Y SET II SENSORS	VER AND
% Max. Power % of Point	Limit Mode	$N_1 = 21,700$ hp = 35,2%	$N_1 = 21, 200$ $hp = 27\%$	$N_1 = 20,700$ hp = 19,7%	$N_1 = 20, 200$ hp = 13, 6%	$N_1 = 19,700$ $hp = 8.6\%$	$N_1 = 24,700$ $hp = 100\%$
3hp 3N <sub>1</sub>	N <sub>1</sub> T <sub>7</sub> W <sub>f</sub>	-6.85 -3.39 -2.24	-7. 28 -4. 15 -2. 81	-7.55 -4.71 -3.18	-7.43 -4.98 -3.31	-5.54 -3.40 -1.56	-3.71 0 0
Ohp QT3	$\begin{array}{c} N \\ T_7 \\ W_f \end{array}$	05 01 + .01	- 10 + 02 0	15 + .04 01	20 + .09 01	25 + . 14 0	000
$\frac{\partial h_{\mathbf{p}}}{\partial \mathbf{w_{\mathbf{f}}}}$	$N_1$ $T_7$ $W_f$	+ .42 +1.03 46	+ .41 +1.03 46	+ .41 +1.04 46	+ .40 1.04 46	+ .40 1.05 46	. 56 0 1.47
$\frac{\partial hp}{\partial T7}$	N <sub>1</sub> T <sub>7</sub> W <sub>f</sub>	+7.70 +3.15 +7.88	+8.90 +4.30 +9.12	+10.02 +5.36 +10.27	+11,73 +6,98 +12,00	+13, 27 +8, 43	0 4.31 0
<del>дьр</del> дт9	$N_1$ $T_7$ $W_f$	-8, 59 -8, 59 -8, 59	-9, 94 -9, 94	-11.29 -11.29 -11.29	-13.04 -13.05 -13.05	-14.75 -14.76 -14.76	000
$\frac{\partial hp}{\partial N_2}$	${f r}_1^{\rm N}$ ${f r}_7^{\rm T}$ ${f W}_{f f}$	+ + + • 93	+1.25 +1.25 +1.25	+1.56 +1.56 +1.56	+1,93 +1,93 +1,93	+2.11 +2.11 +2.11	000

			TABLE X	TABLE XXII - Continued	pe		
Max. Power Limit	Limit Mode	$N_1 = 21,700$ hp = 35,2%	$N_1 = 21, 200$ hp = 27%	$N_1 = 20,700$ $hp = 19,7\%$	$N_1 = 20, 200$ $N_2 = 13, 6\%$	$N_1 = i 9, 700$ hp = 8, 6%	$N_1 = 24,700$ $M_2 = 100\%$
$\frac{\partial}{\partial h}$	z F	+7.59	+8,46	+9, 19	79.97	+9,73	92
Ţ	J <sub>M</sub>	+6.11	+6.38	+8°.79 +7°.07	+9, 65	+9, 45	0
ş	N N	-8.29	-9,17	-10,13	-11.67	-13,19	0
on D	$T_{7}$	-4.49	-5, 25	-5.97	-7.29	-8.52	0
	Wf	-7.28	-8.14	-9.01	-10,45	-11,85	37
ŝ	Z	66 -	- ,52	+ .03	+ .07	+ .10	0
A De	$^{7}$	-1.06	57	90 • -	11	10	0
<u>;</u> p	$M_{\mathbf{f}}$	-1.12	61	12	- ,11	12	0
$\frac{\partial}{\partial N_1} \frac{hp}{SET}$	z	0	0	0	0	0	3,71
θ hp 9T7 SET	T7	0	0	0	0	0	4,31
9	$N_1$	+ .32	+ .30	+ .29	. 28	. 27	78
o iip	$^{7}$	+2,95	+2.97	+2,98	3,02	+3.05	-3, 29
ВГ	$W_{\mathbf{f}}$	+1.02	+1.01	+1.00	1,00	1.00	-1.43
ο hp	z	64	71	92 -	- , 83	88.	. 22
$\langle \overline{\text{SPE}} \rangle$	$T_7$	+1,99	+2,15	+2,31	+2,52	+2,70	-1,57
$\langle w_a \rangle$	J <sub>W</sub>	£ . 69	+ .74	+ .78	+ 83	+ .87	89

			TABLE X	TABLE XXII - Continued	pa		
% Max. Power Limit $N_1 = 21,700 N_1 = 21,200 N_1 = 20,700 N_1 = 20,200 N_1 = 19,700 N_1 = 24,700 N_2 = 24,700 N_3 = 100,700 N_$	Limit Mode	$N_1 = 21,700$ hp = 35,2%	$N_1 = 21,200$ hp = 27%	$N_1 = 20,700$ hp = 19.7%	$N_1 = 20, 200$ hp = 13, 6%	$N_1 = 19,700$ hp = 8,6%	$N_1 = 24,700$ $hp = 100\%$
Ohp OP7	N <sub>1</sub> T <sub>7</sub> Wf	+1.09 + .77 + .96	+ . 22 27 + . 02	59 -1. 23 85	67 -1.62 -1.06	75 -1.98 -1.25	0 0 0
NOTES: $N_2 = 100\%$		standard-day conditions.	nditions.				

TABLE XX.V. ERROR SUMMARY SHEET - PREDICTION AT 35.2% POWER AT  $N_2$  = 100% USING INTERPOLATED "C" IN PREDICTION ALGORITHM SET III SENSORS

Error Source	N <sub>1</sub> Limit % Power	•	Wf Limit % Power	
Linearization of nonlinear				
differential equations	.50	1.23	. 35	
Nonstandard day	.10	. 23	. 21	
Use of base engine "C" for	• = 0	• =0	• =-	
degraded engine	. 03	.50	. 03	
High-power degradation being different from low-power			. 00	
degradation	. 22	.68	. 22	
Change in actual power due to uncertainty in WBL and SPE				
at high power	1.17	4.96	2.15	
Uncertainty in WBL and SPE				
at low power	. 25	1.59	.72	
Sensor errors at low power*	6.40	5.42	6.24	
Control limit and sensor errors				
at high power*	52	1.55	.74	
TOTAL ERROR:	6.56	7.82	$\overline{6.70}$	

63

- EFFECT OF SENSOR ERRORS AT LOW POWER AND AT WER REMAINING ACCURACY SET IV SENSORS	$N_1 = 20, 200 N_1 = 19, 700 N_1 = 24, 700$ $N_2 = 13, 6\%$ $N_3 = 8, 6\%$ $N_3 = 100\%$	-10.92 -12.75 -3.71 -7.19 -8.54 0 -6.38 -8.14 0	24 28 0 + . 13 + . 21 0 02 0 0	6162 .56 +.01 +.02 0 -1.46 -1.46 1.47	+ .83 + .82 0 + .59 + .53 0 + .73 + .71 0	1214 0 -4.09 -4.05 -4.31 +.24 +.26 0	+2,72 +4,52 0 +1,89 +2,78 0 +2,38 +3,80 0
F SENSOR FING ACCUR	$N_1 = 20,700 \text{ N}_1$ $N_2 = 19,7\%$ hp	-9.06 -5.60 -4.49	20 + . 04 04	60 + .01 -1.47	+ .83 + .65 + .76	09 -4.13 + . 22	+1,75 +1,36 +1,59
STUDY - EFFECT OF SENSOR ERRC ON POWER REMAINING ACCURACY	$N_1 = 21,200 N_1$ $p_2 = 27\%$ np	-7.89 -4.41 -3.29	16 01 05	60 + .006 -1.47	+ .87 + .73	07 4.15 +.19	+1,22 +1,02 +1,14
STUDY ON PO	950 N <sub>1</sub> = 21, 700 N <sub>1</sub> 7% hp = 35, 2% hp	-7.09 -3.47 -2.43	12 06 06	60 0 -1.47	+ + + + 83	06 -4.17 +.16	+ + + + 4.88
SENSITIVITY HIGH POWER	= 22,	-5.31 -1.57 96	05 12 05	58 005 -1.47	+ + + + 93	04 -4. 19 + . 08	+ + + + 36
TABLE XXV.	Power Limit N <sub>1</sub>	$N_1$ $T_7$ $W_f$	$^{ m N_1}_{ m T_7}$	$\begin{array}{c} N_1 \\ T_7 \\ W_f \end{array}$	${f r}_1$ ${f r}_7$ ${f W}_f$	N1 T7 Wf	$^{\rm N_1}_{\rm 7}$
TABL	% Max. Power	Ohp On1	Ohp OT3	$\frac{\partial hp}{\partial w_f}$	Ohp Osh P	3hp 3T7	$\frac{\partial hp}{\partial N_2}$

	$00 \text{ N}_1 = 24,700$ hp = $100\%$	76	0	0	0	0	37	0	0	0		3.71	4.31	78	-3, 29	-1.43	. 22	-1,57	89		
	$0 N_1 = 19,700 N_1$ hp = 8,6% hp	+6.29	+6.83	+3.62	888	54	99°+	+ .24	90°+	+ 03		0	0	+ .65	+3,49	+1.41	56	+3.08	+1,24		
	$N_1 = 20, 200 N_1$ hp = 13.6% hp	+5,60	09 <b>*</b> 9+	+3,12	76	80	09. +	+ .24	+ .07	90°+		0	0	<b>19.</b> +	+3,45	+1,41	51	+2.87	+1,17		
- Continued	$N_1 = 20,700 N_1$ hp = 19.7% hp	+4,53	+6.01	+2,28	76	88	+ .55	+ .22	+ .10	+ .07		0	0	69°+	+3,42	+1.42	46	+2.64	+1,10		
TABLE XXV -	$N_1 = 21,200 \text{ N}_1$ hp = 27% hp	+3,94	+5.54	+1.78	70	86	+ .53	+ .20	+ .12	+ .07		0	0	+ .70	+3,40	+1,42	42	+2,46	+1.03	of airflow).	
TA	$N_1 = 21,700 N_1$ hp = 35.2% hp	+3, 52	+5.21	+1.46	89	-1.02	+ .52	+ .22	+ .15	***************************************		0	0	+ .72	+3,38	+1.43	. 38	+2.27	96 +	Units are % max. power/(hp/PPS of airflow). $= 100\%$ standard-day conditions.	
	$N_1 = 22,950 N_1$ hp = 60,7% hp	+2,65	+4.48	96°+	82	-1.26	+ .37	+ .18	+ .16	+ .10		0	0	+ .75	+3,34	+1.43	- ,31	+1.91	+ .82	* Units are % max. power/(hp/PP8 $N_2=100\%$ standard-day conditions.	
	Limit N <sub>1</sub> Mode hp	z	$^{L}$	Wf	N I	$^{\mathrm{T}}$	Wf	z N	$^{L}$	Wf		Z Z	T7	Z Z	$^{\mathrm{L}}$	Wf	z	$T_7$	Wf	its are 100% s	
	% Max. Power Limit NI % of Point Mode hp	o c	£	110		) de	I I	ر د د د د	J. P.	, t	d hn	DNI SET	ahp dry set	- C	Tume	n n	* dye	SPE	$\lceil \langle w_a \rangle \rceil$	NOTES: * Un N2 =	

SING	rror Wf Limit	. 0292	. 6762 1.726 . 0688	. 0109 . 6132 . 0702	1.96 er:	. 715 . 288 . 77
DETAILED ERROR LIST - SENSOR ERRORS AT 35.2% POWER USING SENSORS AT LOW POWER AS TABULATED BEST AVAILABLE SENSORS	hp Prediction Error $ m N_1$ Limit $ m Y_7$ Limit $ m W_f$ Limit	. 0416	0 1.602 1.793	. 0101 2.188 . 1377 . 0215	3.25 n at low pow the Variation it T7 Limit	1.69 .681 1.82
DETAILED ERROR LIST - SENSOR ERRORS AT 35.2% POWER SENSORS AT LOW POWER AS TABULATED BEST AVAILABLE SENSORS		. 0851	. 2760 1. 810 . 0258	. 0114 1.478 . 0918	r 2.34  Ift power extraction  Wf Limit N1 Limit	.36
R ERROR	y Wf Limit	-2.43	-1.47 + .83 + .16	+ .84 +1.46 + .52 -1.08	ror haft power ' ' Wf Limit	+1.43 + .96 cor
T - SENSC VER AS TA	Sensitivity T <sub>7</sub> Limit	-3.47	0 + .77 -4.17	+ .78 +5.21 -1.02 + .15	RSS Error $^{\prime}$ BL) and shaf Sensitivity $^{\prime}$ T $_{7}$ Limit W	+3.38 +2.27 RSS Error
RROR LIS LOW POV	N <sub>1</sub> Limit	-7.09	90	+ + + + + + . 68	ir bleed (W	+.72
DETAILED E SENSORS AT SENSORS	Error in % of Point	+. 012 +. 71	+. 46 +2. 08 +. 43	+. 013 +. 42 +. 135 +. 143	_ ischarge ai	
TABLE XXVI. DE SE SE	Error	+2.5rpm +6.30R	+2.0pph +10hp +6.30R	+2.5rpm +2.2°R +.02psi +.10psi	compressor d	+. 5% +. 3 hp/pps
TABI	Sensor	Low Power: $\frac{N_1}{T_3}$	$\begin{array}{c} \mathrm{w_f} \\ \mathrm{SHP} \\ \mathrm{T}_7 \end{array}$	P H Z Z P J Z Z Z	RSS Error 2.34 3.25 1 Uncertainty in compressor discharge air bleed (WBL) and shaft power extraction at low power:  Sensitivity Variable Uncertainty N1 Limit T7 Limit Wf Limit N1 Limit T7 Limit Wf	$^{\rm W_{BL}}_{\rm SPE}$

TAB	TABLE XXVII.	DETAILED ERROR LIST - SENSOR ERRORS AT 19.7% HORSEPOWER LOW-POWER SENSORS AS TABULATED BEST AVAILABLE SENSORS	ERROR LIS	T - SENSO S AS TABU	R ERRORS	S AT 19.76 SST AVAII	% HORSEPC	DETAILED ERROR LIST - SENSOR ERRORS AT 19.7% HORSEPOWER USING LOW-POWER SENSORS AS TABULATED BEST AVAILABLE SENSORS	
Sensor	Error	Error in % of Point	N <sub>1</sub> Limit	Sensitivity T <sub>7</sub> Limit	W <sub>f</sub> Limit	Z I	hp Prediction Error N <sub>1</sub> Limit T <sub>7</sub> Limit W <sub>f</sub>	rror Wf Limit	
2			(		1				
. T	+2.5rpm	+ 012		0	-4.49	. 1087	. 0072	. 0529	
13 We	+2 Orah	# 4 - I +	07	+ +	04	. 1480	. 0296	. 0236	
SHP	+10hp	+3.62	. + 83.	+ .65	-1.4. + .76	3.005	. 0056 2.353	2.751	
$_{7}$	+6.30R		69 -	4.13	+ .22	. 0414	1.900	. 1012	
$N_2$	$\pm 2.5 \text{rpm}$		+1.75	+1.36	+1.59	. 0228	. 0177	. 0207	
$^{\mathrm{T}_{1}}$	+2.2°R		+4.53	+6.01	+2.28	1.903	2.524	.9576	
P1	$\pm$ . 02psi		92	88	+ .55	. 1026	.1188	. 0743	
P3	$\pm$ . 10psi	+. 17	+ .22	+ .10	+ .07	. 0374	.017	. 0119	
			Tot	Total RSS Error	or	3.58	3.94	3.03	
Uncertainty in compressor discharge air bleed (WBL) and shaft power extraction at low power:	compressor	discharge ai	ir bleed (W	BL) and sh	aft power	extraction	at low pow	er:	
-				Sensitivity		Д	hp Variation		
Variable	Uncertainty	y	N <sub>1</sub> Limit	T <sub>7</sub> Limit	W <sub>f</sub> Limit	N <sub>1</sub> Limit	T <sub>7</sub> Limit	W <sub>f</sub> Limit	
WBL	+.5%		+, 69	+3.42	+1.42	, 345	1.71	.71	
SPE	±.3 hp/pps	SC	46	+5.64	+1,10	. 138	. 792	. 33	
				RSS Error	or	.37	1.88	.77	
									T

TABLE XXVIII. ERROR SUN SHEET - PREDICTION AT 35.2% POWER AT N2 = 10 CONTROL OF SET IV SENSORS

France Course	N <sub>1</sub> Limit	-	•	
Error Source	% Power	% Power	% Power	
Linearization of nonlinear				
differential equations	. 50	1.23	. 35	
$\partial A_N/A_N$ not equal to				
$-\partial \eta_{\rm pt}/\eta_{\rm pt}$	.30	. 54	. 40	
Nonstandard day	.10	. 23	. 21	
Use of base engine "C" for				
degraded engine	.03	. 50	.03	
High-power degradation being				
different from low-power				
degradation	. 22	.68	. 22	
Change in actual power due to				
uncertainty in W <sub>BL</sub> and SPE				
at high power	1.17	4.96	2.15	
Uncertainty in WBL and SPE				
at low power	.38	1.82	.77	
Sensor errors at low power	2.34	3.25	1.96	
Control limit and sensor				
errors at high power	.52	<u>1.55</u>	.74	
TOTAL ERROR	2.77	6.59	3.16	

Since accuracy improves when the predictions are made at higher power levels, errors due to sensor inaccuracies at approximately 60% power were calculated for the system using Set II and Set IV low-power sensors; i.e., T9 or T7 without P7 and the approximation  $\partial A_N/A_N = -\partial \eta_{pt}/\eta_{pt}$ . For single-engine applications, prediction at 60% power would not be feasible since it is desired to make the prediction while still on the ground (30% or less of maximum rotor power). In twin-engine installations, however, one engine could be operating at 60% power while the other engine is essentially at 0% power. Table III also includes sensitivities at 57.5% power. Tables XXIX and XXX list the errors due to sensor inaccuracies at approximately 60% power for Set II and Set IV sensors, respectively. There is no significant difference in the error between the two sets of sensors. Comparing Table XIX to Table XXIX and Table XXVI to Table XXX, there is a moderate improvement in the power prediction accuracy due to low-power sensor errors.

## ALTERNATE MPA SYSTEM STUDIES

The power prediction model described below is presented as a possible alternative approach to predicting maximum power available. This alternate model is significantly less complex than the primary model since it does not involve the rather elaborate gas-path analysis used in the primary model. The prediction concept is based on the assumption that the engine performance remains unchanged after the power prediction calibration. A major disadvantage of this simpler approach is that changes in maximum power caused by degradation in the engine (since power prediction calibration) directly result in a power prediction error.

The alternate power prediction scheme can be divided into two major operations: a simple one-point calibration, and the actual power prediction operation. The calibration operation is done once and thereafter is repeated only when it is ascertained that the engine health has deteriorated from the time the previous calibration was made.

A detailed outline of the alternate power prediction model is given below. The first step establishes a low-power reference value for  $(P_3/\delta_1)$ ,  $(T_7/\theta_1)$ , and  $(W_f/\delta_1 \ \theta_1 \cdot ^{743})$  at a corrected speed of 21,500 rpm; the 21,500-rpm point corresponds approximately to 30% power. The first time that the first step is executed, the values for the above three corrected parameters are stored for future power prediction.

The second step establishes the optimum horsepower corrected to standard-day sea-level conditions at each of the control limits. These values are also stored for future power prediction.

DETAILED ERROR LIST - SENSOR ERRORS AT 60.7% HORSEPOWER USING LOW-POWER SENSORS AS TABULATED BEST AVAILABLE SENSORS	Sensitivity hp Prediction Error $N_1$ Limit $T_7$ Limit $W_f$ Limit $N_1$ Limit $T_7$ Limit $W_f$ Limit	-1.5796 .0584 .0173 .0106 1205 .0335 .0804 .0335 005 -1.47 .203 .0018 .5145	+ .89       + .91       1.0974       1.0502       1.0738         -4.19       + .08       .0160       1.676       .0320         + .34       + .35       .0047       .0044       .0046         +4.48       + .96       1.113       1.882       .4032	-1.26 + .37 .1115 .1714 .0503 + .16 + .10 .0216 .0192 .0120 RSS Error 1.58% 2.74% 1.26%	or discharge air bleed (WBL) and shaft power extraction at low power:  Sensitivity  hp Variation inty  N1 Limit T7 Limit Wf Limit N1 Limit T7 Limit Wf Limit	+3.34 +1.43 .375 1.67 .715 +1.91 + .82 .093 .573 .246 RSS Error .386 1.81 .77
TABLE XXX. DETAILED ERROR LIST LOW-POWER SENSORS	Error in Error % of Point	+1+1+1	+10hp +1.18 +6.3 <sup>0</sup> R +.40 +2.5rpm +.013 +2.2 <sup>0</sup> R +.42		ompressor discharge a Uncertainty	+. 5% +. 3 hp/pps
TABLI	Sensor	$N_1$ $T_3$ $W_{ m f}$	$\begin{array}{c} \mathrm{SHP} \\ \mathrm{T}_7 \\ \mathrm{N}_2 \\ \mathrm{T}_1 \end{array}$	P <sub>1</sub>	Uncertainty in compresson Variable Uncerta	W <sub>BL</sub> SPE

The third step in the outline describes the actual power prediction operation. Each time a power prediction is to be made, the values of  $(P_3/\delta_1)$ ,  $(T_7/\theta_1)$ , and  $(W_f/\delta_1\theta_1\cdot^{743})$  are obtained at a corrected speed of 21,500 rpm as described under the first step of the outline. The present computed values are then compared to the original stored values. If the corresponding values agree to within specified tolerances, then the predicted power obtained under the third step is credible; otherwise it is not. The above tolerances have not been established but would be strongly influenced by sensor inaccuracy and also by anticipated engine degradation.

The constants required in the several equations given for the alternate system are obtained from the steady-state engine characteristics for the base engine. The values of the constants are given in Table XXXI.

- I. Establish Low Power Reference Values for  $(P_3/\delta_1)$ ,  $(T_7/\theta_1)$ , and  $(W_f/\delta_1\theta_1 \cdot 743)$  at  $N_1/\sqrt{\theta_1} = 21,500$  ( $\approx 30\%$  power).
  - A. Measure T<sub>1</sub>
  - B. Compute  $\theta_1 = T_1/518.7$
  - C. Compute N<sub>1</sub> set = 21,  $500\sqrt{\theta_1}$
  - D. Pilot sets actual N<sub>1</sub> speed to a value equal to N<sub>1</sub> SET±200 rpm.
  - E. Allow engine to achieve steady state and measure  $N_1$ ,  $P_3$ ,  $T_7$ , and  $W_f$ .
  - F. Compute:

Note: Step I-F is repeated periodically to correct for engine deterioration,

C. Extrapolate corrected measurements to  $N_1/\sqrt{\theta_1} = 21,500$  thus:

- H. When calibrating, the values obtained under (G) are stored for future power prediction.
- II. Establish optimum horsepower corrected to standard-day sea-level conditions at each control limit.
  - A. At or near max, power, measure the following  $T_1$ ,  $P_1$ ,  $N_1$ ,  $N_2$ ,  $W_f$ ,  $T_7$ , SHP, and  $P_1/P_{AM}$ .
  - B. Compute corrected quantities

$$\theta_{1} = T_{1}/518.7$$

$$\boldsymbol{\delta}_{1} = P_{1}/14.7$$

$$N_{1C} = N_{1}/\sqrt{\theta_{1}}$$

$$N_{2C} = N_{2}/\sqrt{\theta_{1}}$$

$$SHP_{C} = SHP/\boldsymbol{\delta}_{1}\sqrt{\theta_{1}}$$

$$W_{fC} = (W_{f}/\boldsymbol{\delta}_{1}\theta_{1}\cdot^{674})$$

C. Apply correction factor to SHP  $_{C}$  for nonoptimal N $_{2}$ 

$$(N_{2C})_{OPT} = f(N_{1C})$$
 (curve lookup)  

$$CF = \left\{ \left[ N_{2C} - (N_{2C})_{OPT} \right] / (N_{2C})_{OPT} \right\}^{-2}$$

$$SHP_{CO} = SHP_{C}/(1 - CF)$$

where OPT is optimal.

D. Predict N<sub>1</sub> limit (actual rpm)

$$N_{1 \text{Lim}} = 24,700 [(793/W_{fC})] \cdot 09322 (N_{1 \text{C}}/24,700) \cdot 3941$$

E. Compute optimal horsepower at N<sub>1 Lim</sub>, T<sub>7 Lim</sub>, and W<sub>fLim</sub>.

$$\begin{array}{lll} hp_{CNREF} &=& SHP_{CO} \left(N_{1Lim}/N_{1C}\right)^{CHN} \left(P_{1}/P_{AM}\right)^{-C1} \\ hp_{CTREF} &=& SHP_{CO} \left(T_{7Lim}/T_{7C}\right)^{CHT} \left(P_{1}/P_{AM}\right)^{-C2} \\ hp_{CWREF} &=& SHP_{CO} \left(W_{fLim}/W_{fC}\right)^{CHW} \end{array}$$

F. Store Ni Lim, hp<sub>CNREF</sub>, hp<sub>CTREF</sub>, and hp<sub>CWREF</sub> for future power prediction.

Note: Step II is repeated periodically to correct for engine deterioration.

III. Predict Maximum Power Available (MPA).

At some future time after the calibration procedure outlined under steps I and II have been completed, the following procedure is used to predict maximum available horsepower:

- A. Repeat I-A through I-G.
- B. Compare present computed values of  $(P_3/\delta_1)$ ,  $(T_7/\theta_1)$ , and  $(N_F/\delta_1\theta_1\cdot^{743})$  at  $N_1\sqrt{\theta_1}=21,500$  to the stored values. If the values agree within specified tolerances (to be established), then the following power prediction is credible; otherwise, it is not.

On N<sub>1</sub> limit:

SHPON = 
$$(hp_{CNRE F})(\delta_1)(f_N)(P_1/P_{AM})^{C1}$$

On T7 limit:

SHPOT = 
$$(hp_{CTREF})(\delta_1)(f_T)(P_1/P_{AM})^{C2}$$

On Wf limit:

SHPOW = 
$$(hp_{CWREF})(\delta_1^{C3})(f_W)$$

 $f_N$ ,  $f_T$ , and  $f_W$  are defined by Figures 11, 12, and 13, respectively.

$$MPA = MIN (SHPON, SHPOT, SHPOW)$$

# Error Analysis (Alternate Model)

Table XXXII summarizes the error for the alternate power prediction mcdel.

Details regarding the error calculations can be found in Tables XXXIII and XXXIV.

In addition to having large errors in predicting maximum power, this model has the severe disadvantage of not being able to compute changes in maximum power due to engine degradation (since calibration). For these reasons, this model is judged to be unacceptable for a power prediction system.

#### FINAL SYSTEM SELECTION

Based on accuracy considerations and the number of sensors required, the power prediction model selected is the basic power prediction concept with sensor Set IV. The final selected system is described as follows.

At a relatively low power condition not to exceed 30% of the nominal sea-level standard-day value of maximum power turbine horsepower for single-engine applications, the following sleady-state engine and environmental parameters are measured.

# TABLE XXXI. ALTERNATE POWER PREDICTION MODEL DEFINITION OF CONSTANTS

<del></del>		
	Symbol	Value*
	CPNPL	3,1423
	CTNPL	1.235
	CWNLP	4,5782
	CHN	5.5394
	$\mathbf{c}_1$	1.139
:	$c_{HT}$	3.8169
•	$C_2$	1.1078
	$C_{HW}$	1.319
•	$c_3$	8406

<sup>\*</sup>Numerical values were obtained from the analytical baseline engine characteristics.

TABLE XXXII. SUMMARY OF ERRORS FOR THE ALTERNATE PREDICTION SYSTEM	ERRORS FOR	THE ALTI	ERNATE PR	EDICTION SY	STEM
Error Source	N <sub>1</sub> Limit % Power	T <sub>7</sub> Limit % Power	Wf Limit  % Power	Detailed Error List Table No	Notes
Control limit and sensor errors	2		2		
<b>H</b> .	2,35	2,95	2,30	пххх	
Not being at the limit when establishing reference horsepower					
values	462	009	294		1
Nonstandard day	4.52	<b>-6.</b> 50	-3,33		2
Sensor errors at low power	- 477	1,51	. 21	XXXII	က
Degradation since last calibration	. 29	.73	. 20	ШХХХ	
Uncertainty in shaft power and					
bleed air at high power	. 20	.91	.40	HXXXIII	
RSS Error	5, 15	7.41	4,09		
	NOTES	Si			
1. The error resulting from not being at the limit when establishing reference horsepower	sing at the ling	nit when es	tablishing re	ference horse	epower
Values was cotallice by assumme	g unat une mg	m rawod-n	easurements	s can be made	as mucu

- as 500 rpm N<sub>1</sub> below the N<sub>1</sub> limit.
  - Errors due to noustandard-day conditions were computed for the base engine as follows: ત્રાં
    - On N<sub>1</sub> limiting, the horsepower ref. corresponding to a -12°F ambient temperature was computed. This is the lowest temperature that can exist and still be on the N<sub>I</sub> limit.
- On T7 limiting, the horsepower ref. corresponding to a 120°F day was computed. <u>.</u>
  - On Wf limiting, the horsepower ref. corresponding to a -60°F day and  $P_{\rm I}$  = 11.3 psia was computed.
    - At low power, the only sensors which affect power prediction are Pl and T7. <del>.</del>

TAB	TABLE XXXIII.	DETAILED CONTROL POWER	ERROR LI	ST FOR TH SENSOR E	E ALTERNA RRORS AT I	TE PREDI	DETAILED ERROR LIST FOR THE ALTERNATE PREDICTION SYSTEM CONTROL LIMIT AND SENSOR ERRORS AT HIGH POWER AND LOW POWER
High-Power Sensor	Error in % of Point	S N <sub>1</sub> Limit	Sensitivity T <sub>7</sub> Limit	Wf Limit	hp Pr N <sub>1</sub> Limit	hp Prediction Error imit T <sub>7</sub> Limit W <sub>f</sub>	ror Wf Limit
${\rm T_1} \\ {\rm P_1}$	+. 42 +. 143	1.527	3.317	.389	. 641	1.393	. 163 . 046
X W T		-3.36 516 0	0 0 -3.82	0 -1.319 0	. 0336	0 0 1.38	99.
$^{\mathrm{np}}_{1}$ SET $^{\mathrm{N}_{1}}$ SET $^{\mathrm{N}_{7}}$ SET $^{\mathrm{wf}}$ SET	1 + 2 · 0 1 + · 08 1 + · 03 <b>4</b>	1.1 5.539 0 0	1.1 0 3.82 0 RSS	1.1 0 2 0 1.319 RSS Error	2.2 . 443 0 0 2.35%	2.2 0 0 0 2.95%	2.2 0 0 . 045 2.3%
Low-Power Sensor	Error in % of Point	S. N <sub>1</sub> Limit	Sensitivity T <sub>7</sub> Limit	Wf Limit	hp Pr N <sub>1</sub> Limit	hp Prediction Error imit $T_7$ Limit $W_{ m f}$	ror W <sub>f</sub> Limit
$^{\mathrm{P}_{1}}_{\mathrm{1}}$	+. 136 	1 -1.081	1 -3.55 RSS	8406 5 425 RSS Error	. 136 . 457 . 477	. 136 1.5 1.51	.114

T,	TABLE XXXIV.	DETAILED I ERRORS DU ERRORS DU	DETAILED ERROR LIST FOR THE ALTERNATE PREDICTION SYS ERRORS DUE TO DEGRADATION SINCE LAST CALIBRATION AND ERRORS DUE TO UNCERTAINTY IN SPE AND W <sub>BL</sub>	FOR THE ADATION S	ALTERNATINCE LAST	TE PREDIC CALIBRA' WBL	DETAILED ERROR LIST FOR THE ALTERNATE PREDICTION SYSTEM ERRORS DUE TO DEGRADATION SINCE LAST CALIBRATION AND ERRORS DUE TO UNCERTAINTY IN SPE AND $w_{ m BL}$	
Variable	Assumed Degradation	hp N <sub>1</sub>	1 Limit Error	h <b>p</b> T <sub>7</sub>	hp T <sub>7</sub> Limit ," Error	hp v	hp W <sub>f</sub> Limit 'C'' Error	
″c ″t ″ot	. 15% . 1% . 1%	762 979 1.0044	. 1143 . 0979 . 1004	3.3798 4.3485 1.0061	.5070 .4349 .1006	. 8269 1. 1859 1. 007	.1240	
A <sub>N</sub> W <sub>a</sub> A <sub>5</sub>		-1.6937 1.2929 .793 RSS Error	$\begin{array}{c} .1694 \\ .1293 \\ .0793 \\ \end{array}$	2.4311 0597 -1.5644	. 2431 . 0060 . 1564 . 73%	1105 .0727 3507	• • • • •	
Variable	Error*	N <sub>1</sub> Limit	'C'' Exponent T <sub>7</sub> Limit	t W <u>f</u> Limit	Predicted N <sub>1</sub> Limit	Predicted Horsepower Error $_1$ Limit $_1$ Limit $_2$ Limit	er Error W <sub>f</sub> Limit	
SPE	1. 25	. 0024029	01554 3.29 RSS Error	. 006748 1.43	. 19	. 39 . 82 . 91%	. 17 .36 .40%	
*The error at the tim For shaft For bleed	The errors given represent the assumed maximum differences be at the time of calibration and the values at the time of prediction. For shaft power extraction (SPE), units are hp of airflow. For bleed air (WBL), units are percentage of max. airflow.	ent the assumant on (SPE), unit	ed maximumss at the timets are hp of a	difference e of predict airflow. . airflow.	s between tl ion.	ne values o	*The errors given represent the assumed maximum differences between the values of SPE and $w_{BL}$ at the time of calibration and the values at the time of prediction. For shaft power extraction (SPE), units are hp of airflow. For bleed air (WBL), units are percentage of max. airflow.	

Low-power sensor set  $(T_1, P_1, P_3, T_3, N_1, N_2, W_f, SHP, and T_7)$ . The engine will be considered to be operating in steady state if the following three conditions are satisfied:

- 1. N<sub>1</sub> max. N<sub>1</sub> min. within a frame  $\leq 200$  rpm.
- 2. |∆N<sub>1C</sub>| between frames ≤ 100 rpm.
- 3.  $T_{7C}$  max.  $T_{7C}$  min.  $\leq 2.5^{\circ}$ F for 60 seconds.

In the above conditions, a frame of data represents four speed and four temperature measurements in two seconds of measurement time.

From the low-power sensor set defined above, the following set of corrected qualities is computed:

Corrected Sensor Set 
$$(P_3/\delta_1)$$
,  $(N_1/\sqrt{\theta_1})$ ,  $(T_3/\theta_1)$ ,  $(W_f/\delta_1\theta_1^{Y})$ ,  $(SHP/\delta_1\sqrt{\theta_1})$ ,  $(T_7/\theta_1)$ ,  $(N_2/\sqrt{\theta_1})$ 

where

$$\delta_1 = (P_1/14.69)$$
 and  $\theta_1 = (T_1/518.7)$ 

From stored base lines, the base-line values of the corrected parameters are computed at the same corrected sensor value of  $P_3/\delta_1$ . The stored base lines define the nominal steady-state characteristics as a function of  $P_3/\delta_1$  using linear interpolation. Test data from the AIDAPS program showed a significant engine-to-engine variability in the base lines. It is therefore necessary to store a unique base line for each engine rather than a universal base line for a particular model engine to minimize prediction error. At the same corrected  $P_3/\delta_1$  value, a base-line set of corrected engine parameters is obtained. Base Set (N<sub>1</sub>CB, T<sub>3</sub>CB, W<sub>f</sub>CB, SHP<sub>CB</sub>, T<sub>7</sub>CB). The corrected shaft horse-power measurement is then corrected to optimal shaft horsepower due to operation at nonoptimal N<sub>2</sub> speed.

$$CF = [(N_{2C} - N_{2CB})/N_{2CF}]^2$$
  
 $SHP_{CO} = SHP_{C}/(1 - CF)$ 

Relative deviations of the corrected measurements from the base-line data are computed as follows:

$$DN1 = (N_{1C} - N_{1CB})/N_{1CB}$$

$$DT3 = (T_{3C} - T_{3CB})/T_{3CB}$$

$$DWF = (W_{fC} - W_{fCB})/W_{fCB}$$

$$DSHP = (SHP_{CO} - SHP_{CB})/SHP_{CB}$$

$$DT7 = (T_{7C} - T_{7CB})/T_{7CB}$$

Relative deviations in airflow pumping capacity, component efficiencies, and gas generator inlet nozzle effective area are computed from the matrix equation:

The elements of the B-matrix are actually the influence coefficients developed and include the approximation that  $(\partial A_N/A_N) = (-\partial \eta_{pt}/\eta_{pt})$ . A universal B-matrix is used for all engines of the same model number. Since the value of  $P_3/\delta_1$  at which the low-power sensor readings are taken is not known in advance, several values of the B-matrix sufficient to cover the expected range of  $P_3/\delta_1$  will be stored, and linear interpolation with respect to  $P_3/\delta_1$  will be made to arrive at the proper B-matrix.

The low-power sensor measurements are not sufficient to compute variations in all six performance parameters. The relative variation in power turbine inlet nozzle effective area is obtained from the approximate relationship

$$DAN = -DETAPT$$

Maximum power of each of the control limits is computed from the following equations:

On the T7 limit:

SHPOT = 
$$(SHP_{REF}) (\delta_1) (1 + DWA)^{C11} (1 + DETAC)^{C12} (1 + DETAT)^{C13}$$
  
 $(1 + DETAPT)^{C14} (1 + DAS)^{C15} (1 + DAN)^{C16} f_T$ 

On the N1 limit:

SHPON = 
$$(SHP_{REF}) (\delta_1^{C27}) (1 + DWA)^{C21} (1 + DETAC)^{C22} (1 + DETAT)^{C23}$$
  
 $(1 + DETAPT)^{C24} (1 + DAS)^{C25} (1 + DAN)^{C26} f_N$ 

On the Wf limit:

SHPON = 
$$(SHP_{REF}) (\delta_1^{C37}) (1 + DWA)^{C31} (1 + DETAC)^{C32} (1 + DETAT)^{C33}$$
  
 $(1 + DETAPT)^{C34} (1 + DAS)^{C35} (1 + DAN)^{C36} f_W$ 

 $f_{
m T}$ ,  $f_{
m N}$  and  $f_{
m W}$  represent the ambient temperature correction factor and are stored as univariate functions of ambient temperature. These are universal curves valid for any engine of a given model number. The 'C' coefficients are also unique for a particular model engine. An analytical technique developed by Hamilton Standard is used to compute the elements of the 'C' matrix. For each of the 'C' coefficients, two values are stored, and linear interpolation as defined below is used to obtain the actual 'C' coefficient.

Control Limit	Variable Used in the Linear Interpolation
Nl	$\sqrt{\mathrm{T_{1}}}$ or $\sqrt{ heta_{1}}$
$\Gamma_7$	T $_1$ or $ heta_1$
$\mathbf{w_f}$	$P_1\sqrt{T_1}$ or $I_1\sqrt{y_1}$

From the horsepower computed at the three limits, the minimum is chosen as the maximum power available, that is,

# MPA = MIN (SHPOT, SHPON, SHPOW)

From an accuracy consideration alone, the Set II sensors (T9 without P7) have the smallest total error in predicted maximum power (see Table VII). The selected model has only a slightly higher total error; however, it offers a major advantage over Set II sensors in that it requires one less sensor. With Set II sensors, a T9 sensor is required at low power and a T7 sensor at high power; with Set IV, the T7 sensor is used at both low and high power.

In the event that engine diagnostics are desired in addition to maximum power prediction, then with the selected system only one additional sensor would be required (P7) to achieve one of the best-known engine diagnostic algorithms. Conversely, if engine diagnostics are already present, then maximum power prediction system can make use of the available sensors and not require any additional sensors.

For the selected system, the expected accuracy in predicted power at the three power levels considered using the best available sensors is:

Total RSS Error in Predicted Power Including Uncertainty in WBL and SPE

Power Level	N <sub>1</sub> Limiting	T7 Limiting	Wf Limiting
20% power	±3.87%	±6.97%	±3.91%
35% power	±2.77%	±6.59%	±3.16%
60% power	±2.17%	±6.35%	±2.78%

The  $T_7$ ,  $T_1$ , and horsepower sensors at low power alone result in an error in predicted power of nearly 2%. To achieve the best accuracy, it is necessary to eliminate uncertainty in compressor bleed air and shaft horsepower extraction and to make the prediction at the highest practical power level. On single-engine installations, this level would be approximately 30% power. On twin-engine installations, the k/el would be substantially higher if all engines were operated at zero power except the engine on which maximum power was being predicted.

As mentioned previously, a major contribution to the total inaccuracy in predicted power is the uncertainty in compressor bleed air and shaft power extraction. It may be possible through procedural techniques to significantly reduce and essentially eliminate the error due to these two error sources. If these errors are eliminated, then the expected accuracy of the selected power prediction system at the three power levels considered using the best available sensors is:

Expected RSS Error in Predicted Power at No Uncertainty in WBL and SPE

Power Level	N <sub>1</sub> Limiting	T7 Limiting	Wf Limiting
20% power	±3.67%	±4.52%	±3.17%
35% power	±2.48%	±3.94%	±2.18%
60% power	±1.79%	±3.53%	±1.59%

 $T_1$  accuracy may be improved by selecting platinum resistance sensors to have more nearly uniform characteristics and by computer compensation of errors due to nonlinearity. This selective process will increase sensor cost and computer complexity. However, the  $T_1$  sensor plus interface error under these conditions can be reduced from  $\pm 2.2^{\circ}R$  to  $\pm 1.1^{\circ}R$ . Then the prediction error caused by sensor errors at maximum power (shown in Table XVI) is reduced from .52% to .44% when on the  $N_1$  limit. Also, prediction errors caused by the  $T_1$  sensor errors at 35.2% power (Table XXVI) are reduced from 1.478, 2.188, .6132 to .739, 1.094, .307 when on the  $N_1$ ,  $T_7$ ,  $W_1$  limits, respectively. The total expected RSS error in maximum power when predicted from 35% power is summarized below:

# Expected RSS Error in Predicted Power at No Uncertainty in WBL and SPE

#### Also With T<sub>1</sub> Probes Selected for More Uniform Characteristics

Power Level	N <sub>1</sub> Limiting	T <sub>7</sub> Limiting	Wf Limiting
35° power	±2.11°	±3.46°	$\pm 2.12\%$

A comparison of the prediction errors caused by the various sensor errors (Table XXVI at low power and Table XVI at maximum power) with  $T_1$  sensor

errors reduced by selective matching of probes shows that the dominant errors are now caused by the SHP and T7 sensors.

If in addition to selecting T<sub>1</sub> probes for more uniform characteristics the SHP and T<sub>7</sub> sensor error could be reduced by a factor of say 4, then the total expected RSS error when predicted from 35% power would be:

# Expected RSS Error in Predicted Power at No Uncertainty in WBL and SPE

## Also With T1 Probes Selected for More Uniform Characteristics and SHP and T7

# Sensor Errors Reduced by a Factor of 4

Power Level	N <sub>1</sub> Limiting	T7 Limiting	Wf Limiting
35% power	±1.18%	±2.07%	±1.30%

If sensor inaccuracies could be eliminated altogether, the total expected RSS error in maximum power predicted would be significantly improved. When predicted at a 35% power level, the total RSS error in predicted power, assuming no uncertainty in WBL and SPE and perfect sensors at low and high power would be:

# Expected RSS Error in Predicted Power at No Uncertainty in WBL and SPE

#### Also No Sensor Errors at Low and High Powers

Power Level	N <sub>1</sub> Limiting	T7 Limiting	Wf Limiting
35% power	±.70°	±1.00%	$\pm .78\%$

These remaining errors result from a) linearization of the complex nonlinear differential equations base engine (modeling errors), b) using incremental variations from the base engine model for nonstandard-day conditions, c) using incremental variations from the base engine model for engine degradation, and d) high power degradation characteristics at prediction power. It can be concluded from the above analysis that within the ground rules of the present study, a prediction accuracy of at least #1% is not attainable, even with error free sensors. However, significant improvement in accuracy may be achieved on multi-engine helicopters where the MPA prediction can be made at high power levels, and if the single-point ground computation concept is expanded into a "continuous update system" as described in Appendix I paragraph A of this

report. In a continuous update system, the data stored in the MPA computer would be continuously and automatically modified as the engines are operated through their power range under random ambient and flight conditions.

In this manner, the errors due to ambient variations and engine degradation can be virtually eliminated. The pilot could even run individual engines to their power limit during flight to ensure the best possible MPA reading prior to the next take off.

In light of further possible accuracy improvement under slightly different study ground rules, it is recommended (see Recommendations) that further analytical studies be conducted to determine the feasibility of predicting  $^{-1}PA$  within  $\pm 1\%$  accuracy at higher power levels using a continuous update  $typ_{\sim}$  of prediction system.

# SYSTEM MODEL EVALUATION USING AIDAPS FLIGHT TEST DATA

The actual engine test data obtained from the UH-1 ADAPS program was prepared for evolution of the maximum power available (MPA) algorithm. The ADAPS data reduction program was modified to extract the engine data of interest from the raw flight test data. ADAPS data was selected based on its operating range (horsepower, altitude, ambient temperature). This data was then subjected to the following steady-state test constraints:

- 1.  $N_1$  max.  $N_1$  min. within a frame  $\leq 200$  rpm
- 2.  $|\Delta N_{1C}|$  between frames  $\leq \pm 100$  rpm
- 3. TT9C max. TT9C min. ≤ 2.5°F for 60 seconds

One frame of data represented 2 seconds of flight.

Data from 27 flight test runs passed these test constraints. Of these 27 flights, the 3 test flights having the greatest ambient temperature variations were selected, being  $\Delta T_{Amb} = 31^{\circ}F$ ,  $30^{\circ}F$  and  $29^{\circ}F$ . In addition, 4 flights having the greatest change in altitude were selected, being about 11, 200 ft, 9700 ft, 8000 ft, and 8000 ft.

It is noted that the set of three flights having the largest ambient temperature variations is included in the set of four flights having the largest altitude variation. This is to be expected, because the ambient temperature typically

decreases about 3.5°F to 4.0°F per 1000-ft increase in altitude. Therefore, the flights to the highest altitude are most likely to also have the largest temperature variation.

Several flights were made with altitude variations greater than 10,000 ft. However, only the flights referenced above passed the steady-state test constraints. Test pilots were reluctant to fly above 10,000 ft for safety reasons. (The probability of completing a safe autorotation landing during a power climb to high altitudes or while hovering at high altitude is greatly diminished.) For these reasons, no flights were made to 13,000 ft, or above, which would have resulted in a  $\Delta T_{\rm AM} = 51\,^{\circ}{\rm F}$ . Additional tests of three different engines having degraded engine components were evaluated. These three engines contained a degraded compressor, erroded N<sub>1</sub> nozzles (A5 too large), and a degraded turbine respectively. Safety reasons prevented flight tests with degraded engines at power in excess of 1000 horsepower (about 70% of rated power). Therefore, the MPA algorithm was temporarily modified to correspond to engines and controls having a lower maximum power. This temporary modification permitted studies of the algorithm on degraded engines.

The sensors available in the flight test data were not compatible with the final choice of the MPA algorithm. Specifically, the test data did not contain measurements of gas generator turbine discharge temperature (T7). Therefore, the test data could not be used to evaluate the MPA in the T7 limit mode. In addition, a slightly different algorithm must be used (involving the use of T9 instead of T7 measurements at the low-power conditions).

Actual engine test data obtained from the AIDAPS program could not be used to check the validity of the power prediction system selected as the best overall system because of the lack of actual T7 engine data. Instead, the algorithm which uses T9 without P7 and assumes that DAN = -DETAPT was evaluated using actual engine test data. Because the test data does not contain measurements of gas generator turbine discharge temperature (T7), prediction of the horsepower in the T7 limit mode could not be made.

Eight cases were evaluated: four at sea-level standard-day conditions and four at altitude. Table XXXV summarizes the results of the evaluation study. Before discussing each case separately, several general remarks can be made regarding Table XXXIV:

1. The computed engine deteriorations shown in Table I are not to be interpreted as an accurate indication of engine health. They serve only as an intermediate step in the power prediction scheme.

TABLE XXXV. VERIFICATION OF POWER PREDICTION ALGORITHM USING ACTUAL SENSORS SET II  SENSORS SET II  Case 1 Case 2 Case 3 Case 4 Case 5 Case 6 Cass Brand X LE20791 LE18301 LE17376 Brand X LE2 Bas Brand X LE20791 LE18301 LE17376 Brand X LE2 Bas Brand X LE20791 LE18301 LE17376 Brand X Brand X LE2 Bas Alt. (Ft) 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
12 78 79 11 71 10 1 10 10 10 10 10 10 10 10 10 10 10

Computed Engine Deterioration tion  At  NI Limit  Limit  Predicted Actual  Actual  Actual  Refror  Variable  NI  T3  T9  SHP	TABLE XXXV (Continued)	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	SHPON 1412, 806.91 905.6 990.6 1069 1018 1088 1182.  SHPActual 1365, 896.0 853, 948, 1093 1093 1112.  SHPActual 1240, 865.5 954.6 990.4% 1826 1715 1761  SHPActual 1345, 985, 800, 1000,	mPA 1345. 896.0 800. 948. 1093. 1112 III12  r
				Units rpm oR oR hp

- 2. For all cases, shaft power extraction and compressor bleed air were assumed to be 0.
- 3. The "B" and "C" coefficients required in the prediction algorithm were obtained from the analytical base engine as described previously.
- 4. The predicted MPA is the minimum of the predicted horsepower at  $N_1$  limit and the predicted power at the  $W_f$  limit. The actual MPA is the minimum of the actual measured horsepower at  $N_1$  limit and the actual horsepower at the  $W_f$  limit.
- 5. For cases 2, 3 and 4, which evaluate the algorithm against degraded engines, the algorithm was modified to correspond to engines and controls having a base-line maximum power of 1000 hp. This was necessary because, for safety reasons, flight tests with degraded engines were not carried out at powers much greater than 1000 hp (approximately 70% of rated power).

# Case 1

This case attempts to predict MPA at sea-level standard-day conditions for a nondegraded engine of unknown serial number which is identified as "Brand X". The stored base-line values were obtained at a P3 value of 60.36 psia. This is the closest point to 30% power for which computed values of the "B" coefficients were available. For this case, the actual maximum horsepower available is 1345 hp and occurs at the Wf limit. The maximum power as predicted by the algorithm is 1240 hp. The error in predicted power is -7.8%.

# Cases 2, 3, and 4

These cases attempt to predict MPA at sea-level standard-day conditions for degraded engines. The stored steady-state base-line values prior to degradation as well as the actual steady-state engine data for the degraded engines were obtained from actual flight test data at a  $P_3$  value of 60.34 psia for cases 2 and 3 and at a  $P_3$  value of 69.94 psia for case 4. For each case, the values of  $N_1$  and  $W_f$  corresponding to a steady-state base-line horsepower of 1000 hp were found and used to establish the "C" coefficients at high power.

#### Cases 5 and 6

Cases 5 and 6 evaluate the power prediction algorithm at a standard-day 11,000-ft altitude condition for the "Brand X" engine. Case 5 shows the effect of sensor inaccuracy and uncertainty in bleed air and shaft power extraction at high power. Case 6 shows the effect of comparing actual test data against the analytical base line. For case 6, low power measurements were obtained at a corrected  $P_3$  value of 75.23 psia. This corresponds to approximately 30% power at 11,000-ft standard-day conditions. It should be noted that the predicted power at the  $W_f$  limit is so much greater than the predicted power at the  $N_1$  limit that it would never be selected as the minimum. This is as expected since at altitude, fuel flow is not a limiting factor.

# Cases 7 and 8

Cases 7 and 8 evaluate the power prediction scheme at a standard-day 10,000-ft-altitude condition for a different engine. The comments for cases 5 and 6 apply for these cases as well.

#### Results

The AIDAPS flight test data was obtained with normal flight type sensors which were not of the high accuracy required for power prediction. As a result, the accuracy of the prediction algorithm evaluated cannot be established from the flight test data; however, the computed accuracy of the prediction algorithm evaluated is within the limitation of the available data. The available AIDAPS data was also not sufficient to evaluate the power prediction algorithm selected as the overall best prediction scheme.

#### SYSTEM HARDWARE IMPLEMENTATION

System implementation studies drew heavily upon the accuracy studies previously discussed. The complexity of the prediction system selected for implementation studies requires a digital computer and an indicator unit together with associated sensors.

For the purposes of this study, no dependence upon, or interconnection with, any other system not an essential part of existing Army helicopters was considered. It was also assumed that the MPA system, although useful for single-engine helicopters, will more commonly be used with multiengine helicopters. Therefore, a study was made of separate versus integrated computer installations for multiengine applications.

# System Configuration Study

Since one of the prime purposes of the MPA indicating system is improving flight safety, a major consideration of the basic system configuration is that of failure characteristics and possible operational redundancy. No system configuration can be considered single-failure tolerant unless it can provide the MPA of all engines despite a failure of any MPA system part. Information on the MPA of one engine alone on multiengine helicopters is of very little operational use. If redundant MPA systems (a computer and indicator for each engine) were to be true single-failure-tolerant systems, then all computers would require access to all sensors on all engines and all indicators would require access to all computers. Such a system would be more complex and expensive than a single, nonredundant system with only one MPA computer and indicator per helicopter.

Hardware functional built-in-test (BIT) would be used on either type of system to determine if the electronics were functioning properly. BIT would be supplemented in redundant systems by the ability of the pilot to compare the MPA of each engine. Each engine's MPA should not differ from the others by more than a definable amount.

In a single computer installation, BIT must leave negligible chance for a failure producing readings which are not obviously incorrect. A feasible (and typical) method to supplement BIT for greater reliability of single nonredundant systems consists of making a dummy MPA calculation with simulated sensor inputs.

A further safety consideration is sensor isolation where sensors are shared with flight instrumentation. This is important in redundant systems and is critical for a single nonredundant system because a short circuit between sensor inputs which are used for purposes other than MPA calculation could provide incorrect readings for all engines to which they are connected through the MPA system. Such an effect could persist even though the MPA computer power is removed. The necessary isolation can be implemented either externally or internally to the computer. A recommended external method is to design into the flight instrumentation an isolated, conditioned output. This may be a separate retransmitting potentiometer or a semiconductor amplifier. It is also possible to resistively limit faults inside the MPA computer. If this is done, the isolation must be effective for both power—on and power—off conditions, and it must not have any failure modes which cause shorting between inputs from different engines, or be of a nature that gives a nonobvious error. For other sensors

used only for the MPA system, isolation must be sufficient to prevent overloading of sensors or wiring, and preferably to ensure that no computer fault damages the sensor.

It is concluded that since sufficient reliability and safety measures are available, and since significant size, weight, and cost savings can be realized, a single nonredundant MPA system with sensor isolation, BIT, and the capability to perform dummy MPA calculations is the best overall method of im; mentation.

### System Installation

The MPA computer is expected to be too large for installation behind the cockpit instrument panel. Therefore, a separate instrument-type indicator/control unit (I/CU) is recommended for MPA display and for control of the system from the cockpit instrument panel. The computer can be mounted in an electronics bay or other suitable location. A diagram showing the possible appearance of the I/CU and system interconnections is given in Figure 19.

### MPA Display

Since, in multiengine helicopters, the engines are run up separately, only one display is required for MPA. This display can be used sequentially to read out the MPA for each engine and/or to indicate the total MPA for all engines combined. A digital display is preferred for its accuracy, ease of reading, small physical size, and compatibility with digital processing. An electronic display is also desirable for reliability reasons. Many types are available. This application permits the use of solid-state, light-emitting diodes, which offer the smallest physical size and high reliability.

A rotary switch can be used for mode control, having positions for standby, MPA for each engine, and total MPA. The individual engine positions can be used for the convenience of the pilot to indicate the condition of each engine. The display will be blank until the MPA calculation is complete. It is anticipated that the computer will be able to determine from engine measurements which engine is being examined and perform the calculation correctly. Thus, the pilot can use the mode switch for check purposes without disturbing the calculations.

A reset push button is provided so that the pilot can cancel a readout and command a second calculation to observe repeatability for checking purposes. A fault indicator light will be connected to fail-safe circuits which continually monitor

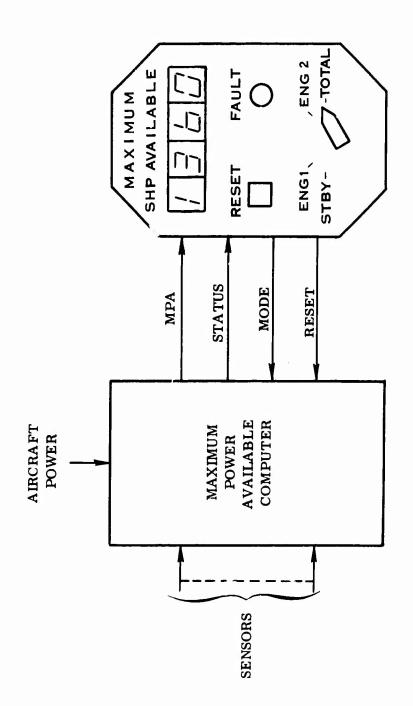


Figure 19. MPA System Configuration.

power circuits and processor capability during a calculation. For an exhaustive check, the processor would perform a dummy MPA calculation in the standby mode using simulated sensor inputs. The result could be displayed and should be visually confirmed by the pilot to within allowable tolerance limits. The computer should also briefly command a lamp test of the display when switched into and out of standby to verify visually that all segments are operating (all 8's display). The fault light should also be tested at this time. The recommended means of communication of the MPA data to the indicator is by a serial digital signal, using a two-wire, twisted, shielded pair. Operational status is shown by sending a ground to the computer, causing the display to be lit from a positive power supply inside the indicator unit. Mode commands are implemented by BCD-coded contact closures to ground, and the system is reset by momentary contact closure to ground in the instrument (reset button).

#### Computer Unit Description

A block diagram of a computer unit is shown in Figure 20. The unit is shown for a twin-engine helicopter. However, this system could easily be expanded into a three-engine capability. The essential sections are a frequency (pulse rate) input interface, an analog input interface, an input interface control, the processor, memory, output interface, and power supply.

### Frequency Input Interface

This section converts frequencies into digital signals. By counting the number of cycles of the input which occur within a fixed time, a basic distinction may be made between measuring the frequency directly and measuring period. In period measurement, the reciprocal of frequency is measured and converted to frequency in the processor. This is the most common method for relatively low frequency inputs, due to the length of time required to accumulate sufficient counts for an accurate frequency measurement. It is necessary to measure the period of several cycles to average out the effects of gear-train and pole-piece irregularities. The normal use of this circuit is signal conditioning of speed inputs and vibrating cylinder type pressure transducers (if used). It will be necessary to multiplex these circuits and to provide a processor programmable divider to accommodate the various frequency ranges. For very low frequencies, where the signal is well below its operating range. the counter may overflow; to detect this, an overflow signal is incorporated to warn the computer to disregard the measurement.

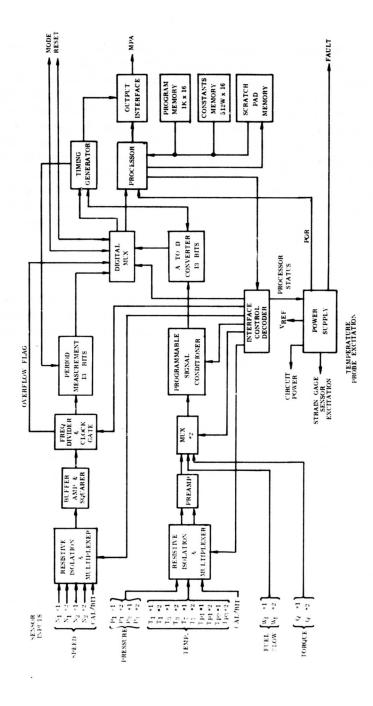


Figure 20. MPA System Block Diagram.

## **Analog Input Interface**

The remaining sensor inputs are dc voltages or voltage ratios. Except for fuel flow and torque, a direct input from the sensor is assumed. The present state of the art permits the input signals to be directly multiplexed without prior individual amplification. Adequate resistive isolation can be provided in each signal lead to prevent overloading or back currents, even under faulty conditions. A preferred form of multiplexing is current multiplexing, in which the computer end of the isolation resistor is grounded when not selected. This eliminates multiplex noise and gives a very accurate, constant input impedance. Additionally, back-to-back diodes may be connected from the computer end of the isolation resistor to ground to give a passive limitation on fault overvoltage.

The pressure sensors used are strain-gage sensors excited by constant voltage. The T<sub>1</sub> temperature sensors are four-wire probes excited by constant current, and the T<sub>3</sub> and T<sub>7</sub> temperature sensors are thermocouples, as are also the correction temperature sensors on the pressure gages. All these signals are less than 1 volt in amplitude and are preamplified before further multiplexing with the 0-5V signals for fuel flow and torque. This second-level multiplexing is followed by a programmable gain amplifier to provide a constant-range signal to the analog to digital converter, thus providing maximum dynamic range and accuracy. A 13-bit converter is required for sensor accuracy considerations.

#### Input Interface Control

A special characteristic of the MPA computer is that although the programmed algorithm is complex, the required operating speed is slow. It is quite reasonable to allow one or more seconds to complete the computer computation. This factor influences interface design and could lead to a special-purpose, simplified processor requiring neither interrupt or direct memory access (DMA) input-output facilities. It is still possible to input data under direct program control, using a combination of program timing, wait loops, and timing flags. The procedure is that the processor will output a control word defining an input channel and gain condition for either of the input interfaces and then repetitively monitor a timing signal until an adequate time has elapsed for signal conversion. In this way, input frame timing is ultimately controlled by the countdown circuits in the crystal-controlled timing generator.

#### Processor

A 16-bit processor has been selected which provides less than .05% of power turbine shaft horsepower round-off error under conservative worstcase conditions. A custom-designed processor having the minimum capabilities for the application can be used to minimize size, weight, and cost and to provide high reliability. The suggested unit could be a microprogrammed serial computer, in which the instruction set is contained within a 1024-word by 8-bit read-only memory. The remainder of the processor would comprise three 16-bit serial-in serial-out registers, a bit counter, a 16-bit inverted shift register for input-output and memory, and a minor amount of additional logic. In low-power off-the-shelf TTL implementation, such a computer requires 16 packages and will perform an addition in 32 microseconds and a division in 3 milliseconds. If implemented in custom MOS LSL it requires two packages and would be 5 or 10 times slower. However, the computation time would still be fast compared with the waiting time for the engine to reach a steady-state condition. The processor has 16 arithmetic and logic instructions and will address directly any memory location or input-output required for the MPA application.

### Memory

Memory is required to store the program instructions, the engine base line, and other curves and constants, and for intermediate calculations (scratch pad). The scratch pad is a read-write memory and is not required to retain data when power is off.

A major implementation consideration is the type of memory to be used for storage of the engine base-line curves. The error analysis has determined that stored engine base-line data must be matched to the particular engine(s) installed. The most operationally desirable base-line acquisition and storage method would be to incorporate the necessary program into the MPA computer itself, and to use an electrically alterable memory for the base-line storage. This could be a plated-wire NDRO memory.

The custom base-line data could be automatically acquired by the MPA computer during a test flight with new (or overhauled) engines installed in the helicopter. A nominal base line could also be permanently stored in read-only memory and used for backup. This technique would ensure continued MPA system operation, although at reduced accuracy, if custom

base-line data could not be acquired because, for example, of helicopter operational problems or in the event of NDRO memory malfunction.

Use of electrically-alterable memory for base-line acquisition and storage is also readily adaptable to an advanced MPA system, where the MPA prediction is automatically updated by the results of previous flights which are stored in the MPA computer. Advanced MPA systems of this nature are discussed in the system growth section of this report.

Since the program has not been reduced to airborne machine language for this study, firm figures for required program memory capacity are not available. A list estimate at this time is 1024 words. In MOS ROM, this would require four packages. The requirement for memory constants is estimated at 519 words for up to three engines, and would require eight packages if implemented in presently available programmable read-only memory. For scratch-pad memory, 64 words are sufficient, which may be obtained in four packages. Scratch-pad memory should be designed to retain data even during short power-supply interruptions.

### **Output Interface**

The output interface receives MPA data words from the processor and encodes them is a form suitable for serial transmission to the indicator/control unit. The transmission is asynchronous and therefore requires a means of indicating the start of each data word. However, no addressing is required as there is only one transmitting unit (the MPA computer), and all required data can be contained in one data word. Error control methods include repetition of transmission, parity, verification of the presence of start and stop bits, and counting of data bits.

The data transmitted is the MPA of each engine or total MPA selected by the mode switch, if available. Otherwise a code is sent, blanking the display. When in standby, the data sent is the result of a self-test calculation, except for one frame which, when changing mode into or out of standby, causes an all-8's lamp test data signal to be transmitted.

### Power Supply

The power supply provides all circuit voltages, excitation for strain-gage and temperature probe sensors, and an accurate reference voltage for the analog-to-digital converters. A primary power source of 115 V, 400 Hz.

single phase is recommended. The power supply checks itself for malfunctions and for power resumption following an interruption. The drive transistor for the fault indicator in the L/CU is located here and is activated either by a power supply fault or by a processor fault, which is a signal resulting from a routine check of processor instructing and addressing performed once per frame. The circuit also has a time-out feature so that program "hang-up" is detected.

In a production configuration, the computer unit might be contained in a volume of approximately 200 cubic inches and might weigh less than 10 pounds. Size and weight are functions of the technology used (e.g., MOS, TTL, etc.) to implement the digital sections of the computer.

### Indicator/Control Unit Description

The I/CU is not required to perform any independent function by itself, and therefore it can be small, simple, and reliable. The only functions requiring electronic circuits are receiving the MPA digital transmission and displaying the data.

The data should be received with a differential line interface circuit, which rejects common-mode interference on the aircraft structure. This data should be verified to ensure that only correctly received data is displayed. It is anticipated that data transmission will be periodic, so that if the first transmission is not correctly received, the data will appear on the second transmission. Display processing includes serial to parallel conversion and BDC to seven-segment conversion. No provision for display dimming should be necessary, as the display is blanked during flight.

The I/CU should contain its own power supply to provide low voltage for the data receive-and-display circuits and for the fault lamp. It is feasible to use 28 volts do if this is preferred by the airframe manufacturer.

In a production configuration, the I/CU might be contained in a standard 2-inch-dia. x 6-inch-deep (less connector) instrument case and might weigh approximately 1 pound.

#### Parametric Sensors

Eleven parametric sensors are required for the MPA system selected as a result of this study. These sensors are listed in Table XXXVI together with their operating range, type and required accuracy. The accuracies presented

# TABLE XXXVI. PARAMETRIC SENSOR SUMMARY

# Sensor Range

Sensor	Min.	Max.	Sensor Type	Sensor Accuracy
NI	20K <b>rp</b> m	25 <b>K r</b> pm	Pulse Pickups	±2,5 rpm
N <sub>2</sub>	13K rpm	22K rpm	Pulse Pickups	±2.5 rpm
Tl	400°R	600°R	Platinum Resistance	±.5°R
Т3	670°R	10 <b>50</b> °R	Platinum Resistance	±.5°R
Т7	1100°R	1800°R	Thermocouple (Chrom-Alumel)	±6.3°R
Т9	1000°R	1500°R	Thermocouple (Chrom-Alumel)	±6.3°R
Pl	ll p <b>si</b> a	16 p <b>si</b> a	Vibrating Cy <b>H</b> nder	±.01 psi
Р3	50 p <b>si</b> a	110 psia	Vibrating Cylinder	±, 05 p <b>ši</b>
P7	20 psia	30 psia	Vibrating Cylinder	±. 03 psi
w <sub>f</sub>	350 PPH	800 PPH	Angular Momentum Mass F/M	±4 pph at 800 pph ±2 pph at 400 pph
SHP	300 hp	1500 hp	Torque Sensor	±10 hp
	300 hp	500 hp		

are achieved via computer compensations to eliminate ali systematic errors associated with each sensor type. In addition, the pressure sensors are individually compensated as well as the torque sensors. Further, improvements in accuracy may be achieved if custom-compensation techniques are used with fuel mass flowmeters, thermocouples and platinum resistance probes. The need for this additional compensation depends on the final overall system configuration. Therefore, individual sensor accuracy requirements can be specified only where the final system configuration and accuracy requirements are defined.

### Speed Sensing

The sensing of engine speeds will be accomplished with the use of magnetic sensors mounted to detect the frequency of passing teeth on shaft-mounted gears. The errors associated with sensing speed are: the tooth-to-tooth position tolerance and the eccentricity of the gears. These errors are very small and are about ±2.5 rpm depending on the particular gear diameter.

#### Temperature Sensing

Platinum Resistance Probes -- Platinum resistance probes were selected for sensing  $T_1$  and  $T_3$  due to excellent repeatability of probe output (resistance) for a given input (temperature). An accuracy of  $\pm$ .5°R can be achieved for the platinum probe. This accuracy of  $\pm$ .5°R is not the normally quoted probe accuracy but is achievable if a digital computer is the termination of the probes signal. The digital computer allows for compensation of the following probe's sources of error:

- non-linearity
- . recovery
- . self-heating
- de-icing

The error sources which cannot be compensated are the probe non-repeatability from unit to unit and the deviation from its calibration with time. The probe can be purchased (at extra cost) with a tighter than normal calibration characteristic, resulting in the total statistical error of unit interchangeability and drift of  $\pm$ . 5°R.

Thermocouple Probes -- The sensing of  $T_7$  and  $T_9$  should be implemented via Chromel-Alumel thermocouples. An accuracy of  $\pm$  6.3°R is required. The maximum limit for using a platinum resistance probe is approximately  $1400^{\circ}$ R. The resultant statistical (RSS) error for a chromel-alumel thermocouple is made up of unit-to-unit variation, and long-term drift. All of the other systematic or repeatable errors are compensated out within the digital computer. These compensatable errors are: non-linearity and recovery.

# Torque Sensing

Torque sensing to an accuracy of  $\pm 10$  hp is required. Such accuracy can be best achieved through the use of a strain-gauge type of sensor. The strain gauge would be mounted on the engine or transmission shaft. The strain gauge output would be a function of shaft torque or horsepower. A temperature sensor would be required to measure the temperature of the shaft, at the strain gauge, for temperature correction of the strain-gauge signal by the digital computer. Also included as compensation by the computer are the gauge nonlinearity and possibly shaft-to-shaft variations. The shaft variations could be compensated on an individual basis by first dead weight calibrating the shaft and including a correction factor in terms of resistors which are part of the shaft-mounted strain-gauge sensor/correction package. Having the compensation resistors as part of the shaft makes the system interchangeable without recalibration.

### Pressure Sensing

The sensing of pressures would be implemented using a highly accurate digital type of pressure transducer. The pressure transducers are accurate to  $\pm$ .06% of full scale under all normal aircraft environment of temperature extremes, vibration, acceleration, etc. These sensors are produced at present in 17, 50 and 100 psi ranges accounting for the .01, .03 and .05 psi errors of the sensor accuracy summary.

The pressure transducer which is able to meet the state accuracies is a vibrating-cylinder device. This transducer has an electrical pulse train (frequency) output which is a function of input or measured pressure. The pressure sensor is composed of a thin-walled sensing cylinder with an end cap on one end and is grounded to a base on the other end (Figure 21). Pressure is ported to the inside of the vibrating cylinder exerting an outward force on the end cap due to the differential pressure across the cylinder and end cap. This outward force results in a tensile load in the cylinder wall as a function of pressure. If the cylinder wall was to be induced into vibration, the frequency of vibration would be a function of the load in the

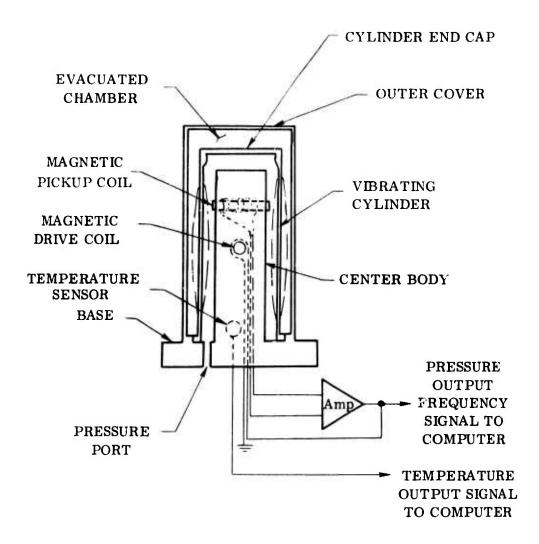


Figure 21. Vibrating Cylinder Pressure Transducer Schematic.

walls, Therefore sensed pressure, similar to a plucked violin string. A closed-loop electronic circuit is added to maintain the wall in vibration at its resonant frequency. A temperature-sensing element is imbedded in the base of the pressure sensor which provides an electrical output as a function of the pressure sensor's temperature. The digital computer then uses the temperature sensors indication of temperature to compensate for the effects of temperature on the pressure sensor's frequency output. Each individual sensor's unique pressure and pressure/temperature characteristics are incorporated on a memory chip mounted with each pressure transducer. This compensation of ambient temperature effects provides for sensor performance which is essentially ambient temperature insensitive.

This vibrating-cylinder pressure sensor is also insensitive to ambient vibration inputs. The range of frequency out of the sensor is 4500 to 5500 Hz which is well above any normal environmental vibration normally specified at a maximum of 2000 Hz.

The remaining sources of error for the pressure sensor are repeatability, long-term drift, and residual temperature effects. The net results of all of these errors are sensors which achieve  $\pm$ .06% of full scale accuracy under all possible aircraft environments.

### Fuel Mass Flow Sensing

Fuel mass flow sensing can be achieved using a linear or angular mass flowmeter. These flowmeters measure mass flow directly and are not volumetric flowmeters with added compensation to do the conversion to mass flow. A temperature probe measuring fuel temperature would still be added to achieve the final state accuracy of  $\pm .5\%$  of the point. The fuel temperature would be utilized by the computer to compensate any residual temperature effects on the flowmeter. The computer will also compensate for any residual nonlinearity of the flowmeters output.

### CONCLUSIONS

Conclusions resulting from this study are as follows:

- 1. The concept of predicting MPA from a helicopter gas turbine engine, while taking into account variations in ambient conditions and engine degradation, is feasible.
- 2. MPA can be predicted with an accuracy of ±3.5% if the prediction is made, under the study ground rules, at 30% power using only the original engine base-line data. Errors associated primarily with parametric sensors, engine degradation and computations made at nonstandard-day conditions and at low-power levels prevent attainment of the ±1% MPA prediction accuracy goal under these study conditions.
- 3. Significant improvement in MPA prediction accuracy can be achieved if the prediction is made at higher power levels, on multi-engine helicopters for example, and if engine base-line data stored in the MPA system is continuously updated. MPA prediction accuracy of ±1% may be feasible under these conditions where errors due to power setting, ambient variations and engine and parametric sensor degradation have been virtually eliminated.
- 4. Parametric sensor requirements have been determined for an MPA prediction system configured under the present study ground rules. Errors associated with power turbine inlet temperature and shaft horsepower sensing are major contributors to MPA prediction error. Sensor accuracy requirements are dependent on system configuration and error compensation techniques. Therefore, final sensor requirements for a continuous updating MPA prediction system with an accuracy of ±1% cannot be determined until such a system is configured.
- 5. An MPA prediction system can be implemented using present-day state-of-the-art digital electronic technology.
- 6. MPA prediction techniques are highly compatible with other diagnostic and control functions. Overall hardware utility can be improved by combining these functions on advanced Army helicopters.

#### RECOMMENDATIONS

Recommendations resulting from this study are as follows:

### 1. Analytical Studies

Analytical studies should be conducted to determine the feasibility of predicting MPA to within  $\pm 1\%$  accuracy at power levels higher than 30% using a continuous update type of prediction system, and to determine the required accuracies of parametric sensors associated with the system.

### 2. Hardware Fabrication

If at the conclusion of these additional feasibility studies, it is concluded that MPA can be predicted with suitable accuracy, then system hardware should be fabricated. A breadboard MPA system could be used in conjunction with a Lycoming T53 engine or engine computer model to verify the analytical findings of the feasibility studies and to optimize system hardware implementation.

Breadboard hardware can be fabricated economically and within a reasonable length of time. Existing state-of-the-art parametric sensors can be used for initial system development.

The breadboard hardware could be designed so that only minor modifications are required to allow the use of more accurate sensors, as they become available, and to allow growth from an MPA system to a hover-lift computer system or small diagnostic system.

At the conclusion of the breadboard evaluation program, a prototype MPA system could be fabricated for hardware development, and finalization under actual flight conditions.

# 3. Advanced MPA System Studies

It was concluded that MPA prediction techniques are highly compatible with other helicopter diagnostic and control functions. It is, therefore, recommended that further studies, such as those described briefly in Appendix I of this report, should be undertaken to determine how improvements in overall hardware utility can be made on advanced Army helicopters by combining these diagnostic and control functions with an advanced MPA prediction system.

#### APPENDIX I

### NOTES ASSOCIATED WITH ERROR SUMMARY SHEETS

### NOTE 1

Linearization of the nonlinear engine equations introduces an error in both "B" and "C". Accuracy checks of individual elements in "B" and "C" indicate that errors of as much as  $\pm 5\%$  may occur. It is assumed that there is a possible  $\pm 5\%$  error in "C" and in the computed degradation at low power (efficiencies and geometrics).

### NOTE 2

The influence coefficients, "C", were computed at standard day ( $P_1 = 14.7$  psia,  $T_1 = 59$ °F,  $P_1/P_{AM} = 1.0$ ). Nonstandard day was assumed to be any combination of the following:

 $P_1 = 11 \text{ to } 16 \text{ psia}$ 

 $T_1 = -60^{\circ} F \text{ to } 120^{\circ} F$ 

Ambient conditions where each limit can occur are defined in Figure 18.

#### NOTE 3

Tolerances of new engines are dependent on both quality and size. Similarly, degradation tolerances are not certain. The assumed engine tolerances used in this study are shown in Table XXXVII.

### NOTE 4

"C" is computed at standard day when using constant "C". Interpolation of "C" based on ambient-day conditions can reduce the error in "C" to less than  $\pm$ . 01 on N<sub>1</sub> limit and W<sub>f</sub> limit and  $\pm$ . 02 error in "C" on the T<sub>7</sub> limit.

### NOTE 5

The variation in "C" caused by a "typical degradation" was evaluated at the condition shown below:

TABLE XXXVII. ASSUMED ENGINE FOLERANCES						
Engine Parameter	New Engine Tolerance	Severe Degradation	Maximum Tolerance			
$\eta_{ m c}$	±.5%	<b>-3</b> %	3.5%			
$\eta_{ m t}$	±.5%	-2%	2.5%			
$oldsymbol{\eta}_{ ext{pt}}$	±.5%	-2%	2.5%			
<b>A</b> <sub>5</sub>	$\pm_{ullet}2\%$	+2%	2.2%			
$A_N$	±. 2%	+2%	2.2%			
$\mathbf{w_a}$	±1%	-1%	2%			
			<u></u>			

Engine Parameter	Typical Degradation
$\eta_{\mathrm{c}}$	-1%
${m \eta}_{ m t}$	-1%
$\pmb{\eta}_{ ext{pt}}$	-1%
A <sub>5</sub>	+1%
$A_N$	+1%

The variation in "C" caused by a "severe degradation" is assumed to be twice the variation in "C" caused by the above "typical degradation."

### NOTE 6

The prediction algorithm assumes that the degradation at high power (in terms of percentage of point  $\eta_c$ ,  $\eta_t$ ,  $\eta_{pt}$ ,  $A_5$ ,  $A_N$ ,  $W_a$ ) is equal to the degradation computed at low power. The estimated errors in this assumption are as follows:

$$\eta_{c} = \pm .15\%$$
 $\eta_{t} = \pm .10\%$ 
 $\eta_{pt} = \pm .10\%$ 
 $A_{5} = 0\%$ 
 $A_{N} = 0\%$ 
 $W_{a} = \pm .10\%$ 

### NOTE 7

Compressor bleed air (WBL) and compressor shaft power extraction (SPE) influence maximum power available. The estimated maximum  $W_{BL} = 5\%$  and maximum SPE = 1 hp/PPS; however, the actual values are dependent on accessory requirements. The assumed uncertainty in predicting high power  $W_{BL}$  and SPE while at low power is shown below:

$$W_{BL} = \pm 1.5\%$$

$$SPE = \pm .3\% \text{ hp.}'PPS$$

### NOTE 8

The compressor bleed air (W<sub>BL</sub>) and compressor shaft power extraction (SPE) are not measured. It is assumed that W<sub>BL</sub> and SPE can be estimated to  $\pm$ .5% of W<sub>a</sub> and  $\pm$ .3% hp/pps of W<sub>a</sub> respectively at low power.

### NOTE 9

The sensor accuracies are considered the "best available." Detailed sensor characteristics are listed in Table XV.

### NOTE 10

The sensor accuracies are typically good sensors, although often not the "best available." In particular, the pressure sensors and shaft horsepower sensor (SHP) are not the "best available." Detailed sensor characteristics are listed in Table XV.

## NOTE 11

It is assumed that  $\partial A_N/A_N$  differs by 2% from the assumed condition of  $\partial A_N/A_N = -\partial \eta_{pt}/\eta_{pt}$ .

#### APPENDIX II

### SYSTEM GROWTH

There are three aspects of possible system growth which should be considered in further studies. They are as follows:

- Modification or expansion of the base-line MPA system to improve accuracy.
- · Incorporation of other aircraft-related functions to attain a hover lift computer.
- · Integration of the MPA function with other aircraft systems.

### Growth To Improve Accuracy

The system implemented is that which complies with the requirements as specified by the Army, both in writing and verbally. A major restriction of these requirements is that the MPA calculation be made only on ground prior to flight.

An alternate approach which would provide an improved accuracy at some increase in complexity is to continually update the MPA calculation in flight whenever the engine operating condition enables a better estimate than that currently made. The decision whether or not to update the reading at any time would be an automatic computer function based on evaluation of the relative merit of a new reading compared to the old. This would be a function of engine and flight condition, age of previous MPA computation, etc.

The improved accuracy of this approach results from the ability to make the MPA calculation at high power levels and the expanded operational envelope, which would also allow evaluation to be made of the engine's deviation from the standard base line. The MPA reading presented to the pilot prior to flight would be that obtained from the last update on the previous flight.

The basic hardware required to implement this approach would be the use of a long-term, nonvolatile read/write memory system such as plated wire. An alternative could be regular scratchpad memory with a mechanically latched readout system.

Other means to improve accuracy without significant hardware impact involves operational procedures in multiengine helicopters. It is, for example, possible in these applications to operate individual engines to higher percent powers without approaching lift-off. The on-ground MPA calculation accuracy is then improved by virtue of the higher power setting.

### Growth to Hover Lift Computer

The MPA indicator as described in this study provides a useful indication to the pilot as to the engine's capabilities. It falls short, however, in that it does not provide him with information pertinent to the performance capabilities of the helicopter itself. It is still necessary to manually assess whether he could lover at a specific altitude to perform his mission, etc.

The hover lift computer could provide this information automatically. The helicopter-related performance characteristics would be incorporated into the computer logic and other related variables would be provided as additional inputs. The most important (and possibly the only) one of these is weight. This could be obtained as dynamic weight (rotor shaft lift), on-ground weight (strain gage on struts), payload capability, etc., and could include the use of fuel flow measurements for weight update. The output information to the crew could be presented in a variety of ways to the pilot. Possibilities range from a go no-go response to interrogation to a flight envelope display.

It may also be required to provide means to manually input the computer with ambient conditions known to exist at the mission destination to enable advance estimates to be made as to whether it will be possible to successfully complete the mission.

#### Integration With Other Systems

A side benefit of the MPA system is that the sensor signals which are conditioned and read into the computer could also be provided to other on-board systems. These include crew instrumentation, diagnostics (including TREND analysis), yaw dampers, and automatic stabilization equipment (ASE). The conditioned signals are most readily made available in a digital format, requiring only coding and outward transmission serially on a single pair of wires. Under some circumstances an analog-type buffered signal could be provided. The accuracy requirement for the MPA system will be the pacing item for the sensors, and it may well be possible to tolerate some accuracy degradation in the instrument

drive channel. The MPA system could also be totally integrated with other on-board systems to improve overall hardware utility, since MPA prediction techniques are highly compatible with diagnostic and control functions on advanced Army helicopters.