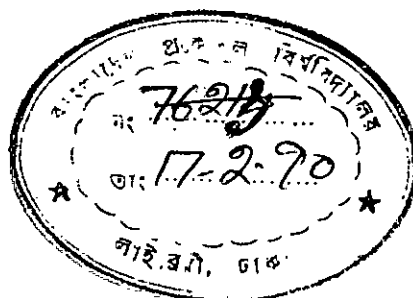


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DEVELOPMENT OF A STATE ESTIMATOR FOR BPDB
(BANGLADESH POWER DEVELOPMENT BOARD) GRID SYSTEM

A THESIS

SUBMITTED TO THE DEPARTMENT OF ELECTRICAL AND ELECTRONIC
ENGINEERING, BANGLADESH UNIVERSITY OF ENGINEERING AND TECHNOLOGY
IN PARTIAL FULFILMENT OF THE REQUIREMENTS FOR THE DEGREE OF
MASTER OF SCIENCE IN ENGINEERING (ELECTRICAL AND ELECTRONIC)

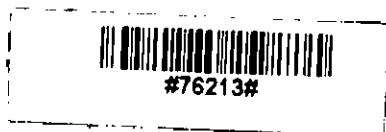


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BY

MD. SHAHJAHAN

November, 1989



DEPARTMENT OF ELECTRICAL AND ELECTRONIC ENGINEERING
BANGLADESH UNIVERSITY OF ENGINEERING AND TECHNOLOGY
DHAKA, BANGLADESH.

CERTIFICATE

This is to certify that this work was done by me and it has not been submitted elsewhere for the award of any degree or diploma.

Countersigned

Signature of the Candidate

S. Shahnawaz Ahmed
(Dr. S. Shahnawaz Ahmed)

Supervisor

shahjahan 15-11-89
(Md. Shahjahan)

The thesis titled
Development of a State Estimator for BPDB
(Bangladesh Power Development Board) Grid System

by
Md. Shahjahan
Roll. 871331F

has been accepted as satisfactory in partial fulfilment of the requirements for the degree of Master of Science in Engineering (Electrical and Electronic).

Board of Examiners:

- (i) S. Shahnawaz Ahmed 25.11.1989
(Dr. S. Shahnawaz Ahmed) Chairman
(Supervisor)
Assistant Professor
Department of Electrical and
Electronic Engineering
BUET.
- (ii) M. G. Matin 25.11.1989
(Dr. Md. Abdul Matin) Member
Professor and Head
Department of Electrical and
Electronic Engineering
BUET.
- (iii) A. Rahman 25.11.89
(Dr. Syed Fazl-i Rahman) Member
Professor
Department of Electrical and
Electronic Engineering
BUET.
- (iv) Dr. Shahidul Islam Khan 20/11/89
(Dr. Shahidul Islam Khan) Member
Associate Professor
Department of Electrical and
Electronic Engineering
BUET.
- (v) Dr. Md. Sekendar Ali 25/11/89
(Dr. Md. Sekendar Ali) Member
Director
(External)
Bangladesh Institute of Technology, Dhaka
Gazipur.

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ABSTRACT

Bangladesh Power Development Board (BPDB) is considering implementation of a computer aided monitoring and control system for its grid network. With this objective ahead it has already installed telemetering instruments at some of its grid substations. The measurements on line flow, nodal injections and voltages acquired through telemetry are inherently error corrupted. To obtain a reliable and consistent set of information on system state from the raw telemetered data an algorithm known as State Estimator is required.

In the present research work a computationally efficient program has been developed for a fast decoupled bad data suppression state estimation algorithm. In this algorithm the usual and random measurement errors have been minimized in least square sense using weighting factors chosen on the basis of statistical properties of the errors. The gross measurement errors have been suppressed using a reweighting technique. The measurements have been grouped into active and reactive sets to take advantage of decoupled characteristics of a high voltage power transmission system. Sparsity was exploited in computations involving matrices.

The developed program has been tested extensively on

BPDB grid system choosing a number of measurement configurations and simulating various sets of errors. The Monte Carlo technique was adopted for simulation test. The results of the comprehensive study on robustness, accuracy of estimates and bad data processing capability of the algorithm have been analysed using established performance indices.

LIST OF PRINCIPAL SYMBOLS

J	Objective function
M	Number of measurements
N	Number of state variables
WLS	Weighted Least Square
LP	Linear Programming
BDS	Bad Data Suppression
[Z]	Measurement Vector
[ΔZ]	Measurement residual vector
[X]	State vector
[\hat{X}]	Estimate of state vector
[V]	Voltage magnitude vector
[θ]	Phase angle vector
[ξ]	Measurement-error vector
[R]	Measurement error covariance matrix
σ	Measurement error standard deviation
[h(.)]	Vector of functions of (.)
[H]	Jacobian matrix
[G]	Gain matrix

Superscripts

i	iteration counter
T	transpose
*	conjugate of a complex quantity

Abbreviations

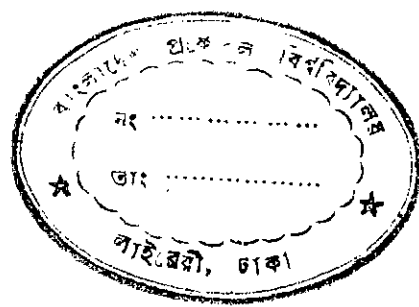
BPDB	Bangladesh Power Development Board
BUET	Bangladesh University of Engineering and Technology
IBM	International Business Machines
CPU	Central Processing Unit

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CHAPTER 1
INTRODUCTION



1.1 General Considerations

A growing demand for electrical energy has constrained the present day power utilities in both developing and developed countries with the vital problem of ensuring an uninterrupted and low-cost delivery of power with negligible fluctuation in the frequency and in the voltage. A basic prerequisite for such a secure and economic operation of a power system is frequent and quick monitoring of the current state of operation. In order to help the operator for monitoring the system operator and then take decision for appropriate control action, a large amount of on-line data is acquired from the system and telemetered to the dispatch centre. The data mainly consists of measured values of line power flows, nodal injections and voltage magnitudes at various nodes of the system. The raw telemetered data is corrupted by inherent random measurement noise and also by large or gross errors. The random noise mainly results from current and potential transformer errors, meter inaccuracies, analogue to digital conversion error and noise in communication channels or interference noise. Gross errors, also called bad data, are the outcome of faults or failures in metering and data communication systems which may also give rise to the situation of missing data. Obviously an algorithm is required to process the raw data to compensate for its various uncertainties and produce a consistent and reliable set of information on the current operating state of the system. The algorithm used for this purpose is called the State Estimator¹.

1.2 Approach of State Estimation

A state estimator is a computer software whose input data is a redundant set of measured quantities and the output is the best possible estimate of the state variables i.e. busbar voltages and phase angles which are used to obtain the computed values of measured, unmeasured and missing quantities. Redundant means that the number of measurements is more than the number of state variables. For the best estimate the state estimator uses a suitable statistical criterion, for instance sum of weighted squares of the difference between observed values of the measurements and their computed values, with a view to minimising the effect of errors in the telemetered measurements. An iterative procedure is involved in state estimation algorithm. In the first iteration the computed values may be obtained from a flat start i.e. unity voltage magnitudes and zero phase angle.

The mathematical model of the system network required for state estimator is formed on the basis of the telemetered status of circuit breakers at different nodes of the system. State estimation is usually performed at prespecified time periods of a daily routine or whenever a major change in the system operating conditions or network configuration takes place.

State estimation differs from a load flow analysis in the following way.

- i) While state estimation processes a redundant set of on-line measurements on injections, line flows and bus voltages and compensates for the errors in them, load flow has nothing to do with measurements. Load flow requires that

the magnitudes of a set of only nodal injections equal to the number of nodes be available and does not care for the uncertainties, source or reliability of the input data so that the results of load flow calculation can be seriously affected.

- ii) In on-line and real-time control²⁻⁵ of a power system, the values of injections are computed more accurately using the estimates of bus voltage and phase angles obtained from the state estimator and then made available for performing a load flow. From this point of view load flow is a secondary outcome of state estimation.

1.3 Applications of State Estimation

The output of the state estimator is accessed by other on-line functions⁶ like security control and the economic dispatch, through a data base. Security control requires reliable information on the system's present state to predict in a very short time i.e. real time the effects of contingencies such as generator tripping, line outage, fault occurrence, loss of load etc. upon the customers demand, nodal voltage limits, generator loading limits, transmission line thermal limits, minimum spinning margin and so on. Economic dispatch function determines the best allocation of generation requirements among the available generating units so that the demanded load can be satisfied within the above mentioned constraints imposed by security considerations at a mini-

mum cost. To this end an optimal power flow problem requiring accurate information on system state, particularly on bus injections is involved. State estimator produces the best estimates of them by using the estimated bus voltages and phase angles.

The state estimator also identifies the grossly erroneous measurements and hence can provide the locations of suspect measurements so that remedial actions can be taken quickly in the field on the faulty components of instrumentation and data communication system.

1.4 Background and Purpose of the Present Work

It is almost two decades since the state estimation theory has been first reported¹ for application in power systems. Mainly three approaches - static, tracking and dynamic state estimation were proposed in the literature.^{1,7,8} The first one uses only a single snapshot of measurements to estimate the states of a system at a particular instant of time. The last two consider the time variation of system state in addition to a particular measurement snapshot. Yet the method of static state estimation is preferred⁹ in practice because of its simplicity in modelling compared to the other two.

For the method of static state estimation two basic algorithms have been proposed so far. The first suggested^{1,9} one is based on minimisation of a weighted least square (WLS) criterion. Later research work¹⁰ has suggested another algorithm based on a linear programming (LP) technique to minimise a linear criterion.

Although the LP based state estimator was found to show better performance in processing bad data compared to the basic WLS estimator, the former requires¹¹ more computer time and storage than the latter.

A number of research work devoted to further improving the computational as well as bad data performance of basic WLS estimators gave rise to three versions namely Fast Decoupled¹² WLS, Line-Only¹³ WLS and Bad Data Suppression^{11,14} (BDS) algorithms. The fast decoupled one takes advantage of the well known weak coupling among active and reactive quantities of a power system and makes a grouping of the measurements into active and reactive sets. Eventually a decoupled algorithm with constant Jacobian matrices incurring less computer storage and time per iteration can be derived. The line-only algorithm was developed only to process a measurement set consisting of only complex line flows. The Bad Data Suppression algorithm was developed by modifying the WLS criterion to assign less weight to the suspect measurements according to an efficient non-quadratic criterion e.g. quadratic square root criterion. The suppression technique can be incorporated in any WLS based estimator at the cost of a trivial change in computer program, time and storage requirements. In general, the fast decoupled WLS estimator is applicable for a metering configuration comprising all types of measurements and when incorporates BDS technique it has been reported¹¹ as the most efficient algorithm in terms of both computational and bad data performance.

In Bangladesh its only public utility - Bangladesh power Development Board (BPDB) has just stepped into the process of

computerizing its central load despatch centre (CLDC) with the installation of telemetering system at some selected grid substation . In course of phase by phase implementation of the project a computer is expected to be installed at CLDC by 1991. This has become an incentive for the researches at BUET to investigate into different aspects of state estimation and make an initial assessment of its applicability for the monitoring and control of the national grid system. Part of the task is the object of the present research work.

The main areas in which the present work contributes are as follows :

- 1) A state-of-the-art review of state estimation algorithms has been carried out in order to choose which one would be the most suitable.
- 2) A computer program has been developed for the WLS state estimation algorithm incorporating fast decoupling and bad data suppression techniques. Also a highly efficient sparsity¹⁵ exploitation program has been interfaced with it.
- 3) The developed program has been tested extensively on BPDB grid system using Monte Carlo simulation¹⁶ technique.

1.5 Organisation of the Thesis

The presentation of the material studied in the research work is organised as follows.

Chapter 2 describes the WLS state estimation algorithm together with fast decoupling and bad data suppression techniques.

Chapter 3 presents the programming details of the fast decoupled BDS algorithm.

Chapter 4 presents the results of simulation tests of the developed computer program.

Chapter 5 provides a summary of the main results obtained in the present research work and suggest some areas for further research.

The appendices include supporting materials to different chapters of the thesis.

CHAPTER 2

METHOD OF FAST DECOUPLED
WLS STATIC STATE ESTIMATION

2.1 Introduction

The weighted least square (WLS) algorithm⁹ estimates the state variables by minimizing an objective function which is the sum of squares of a number of terms with each term corresponding to a measurement acquired from the system and weighted by a factor chosen on the basis of statistical properties of the measurement error. The minimization of the objective function is formulated as an iterative process based on the Newton-Raphson technique which involves the inversion of a matrix known as gain matrix to update the state variables in each iteration. The iterative process stops when the change in each of the state variables between two successive iterations is less than some prespecified value termed as tolerance margin.

The WLS algorithm in its basic form involves a large amount of computer storage and excessive computing time due to evaluation and inversion of the gain matrix in each iteration. These difficulties have been overcome in another version of WLS algorithm known as fast Decoupled estimator¹² taking advantage of decoupling characteristics and a number of simplifications based on physical properties of high voltage power transmission system and engineering judgement.

The measurements to be processed by a state estimator are in general corrupted by usual errors called measurement noise which are of random nature and reported¹⁷ to be governed by

Gaussian¹⁸ (normal) probability distribution law. This noise is inherently filtered i.e. compensated by a WLS estimator. Apart from the noise, for a number of reasons mainly malfunctioning or failure of transducers and telemetry, one or more data in the measurement input can be bad containing gross errors which do not follow the Gaussian law. In order to process gross measurement error effectively WLS estimators are modified as Bad Data Suppression (BDS) algorithm¹⁴ in which only the suspect measurements are reweighted to suppress their effects. The quadratic square root¹⁴ criterion is used for reweighting.

It is necessary to check whether the available number of measurements in a system would be enough to determine its state variables. This possibility is known as observability which depends upon the network as well as metering configurations. Observability of a system needs to be determined prior to performing state estimation and remains valid as long as the network and the measurement configurations remain unchanged.

2.2 Basic WLS Static State Estimator^{1,9}

The method of static estimation estimates the state variables from a single snapshot of measurements taken at a specified instant of time. A simple model relating the measurement (nodal injections, line flows and voltage magnitudes) to the state variables and network parameters is given by

$$[Z] = [h(x)] + [\xi] \quad (2.1)$$

where

[Z] - MX1 vector giving observed (telemetered) values of measurements

[X] - NX1 state vector comprising state variables

[h(x)] - MX1 vector of nonlinear functions in state variables providing the true but unknown values of the measured quantities. These functions are shown in section A.1 of Appendix-A.

x - an element of state vector [X]

[ξ] - MX1 vector of measurement errors.

M - number of measurements

N - number of state variables, i.e. voltage magnitudes and phase angles to be estimated.

$M/N > 1$ - redundancy ratio

The weighted least square algorithm uses this model and minimizes the sum of the squares of the difference between the observed values and true values of the measurements, with each term weighted by a factor w . Equation (2.2) shows the WLS objective function.

$$J(x) = \sum_{m=1}^M w_m (z_m - h_m(x))^2 \quad (2.2)$$

where m denotes measurement serial.

The successive linearisation and minimization of equation (2.2), as shown in section A.2 of Appendix A, leads to an expression for the optimum estimate of the state vector as in equation (2.3).

$$\hat{[X]}^{i+1} = \hat{[X]}^i + [G]^{-1} [H]^T [R]^{-1} [\Delta Z]^{i+1} \quad (2.3)$$

where

$$\begin{aligned} \hat{[X]} & - \text{estimate of state vector} \\ i & - \text{iteration counter} \\ [G] & = [H]^T [R]^{-1} [H] \end{aligned} \quad (2.4)$$

is gain matrix of dimension NXN

$$[H] = \left. \frac{\partial h}{\partial x} \right|_{x=\hat{x}^i} \quad (2.5)$$

is Jacobian matrix of dimension MXN

[R] - measurement error covariance matrix of dimension MXM which is diagonal with constant elements as shown in section 2.2.1

T as superscript denotes transpose

$$[\Delta Z]^{i+1} = [Z] - [h(\hat{x}^i)] \quad (2.6)$$

is measurement residual vector.

If all the measured quantities are linear functions of state variables, e.g. nodal currents and branch currents, equation (2.3) reduces to a simple non-iterative form as in equation (2.7)

$$\hat{[X]} = [G]^{-1} [H]^T [R]^{-1} [Z] \quad (2.7)$$

where all the matrices are with constant elements i.e. independent of state variables. A linear case of the state estimation problem is not the practical one. Because in a power system measuring the

nodal power injections and line power flows (which are nonlinear functions of state variables) is the usual practice and also provides an easier means for estimating the unconventional quantities like phase angles of the buses.

However an example using a simple d.c. network with nodal current and branch current measurements has been produced in section 2.2.2 just to show the basic principle of the WLS state estimation algorithm.

2.2.1 Weighting Factors

The measurement errors represented by the vector $[\xi]$ in equation (2.1) are mainly due to current and potential transformers' errors, inaccuracies of metering instruments (transducers), errors of analogue-to-digital (A/D) converters, noise in communication channels or interference noise. Consequently these errors are of random nature and termed measurement noise. The best measure of the uncertainty associated with any particular measurement is the standard deviation of its error. The weighting factors w used in equation (2.2) can be computed in terms of the standard deviations by assigning a probability law to the measurement noise. It has been reported¹⁷ that the measurement noise usually follows the normal (Gaussian)¹⁸ distribution with zero mean and a known standard deviation σ , and is independent of another measurement noise. Equations (2.8) and (2.9) respectively show these statistical properties of the noise of the m -th measurement.

$$E(\xi_m) = 0 \quad (2.8)$$

$$E(\xi_m \xi_m^T) = \sigma_m^2 \quad (2.9)$$

where,

E - statistical expectation operator

σ_m - standard deviation of the noise of m -th measurement.

The smaller the standard deviation the greater is the confidence which can be placed in the observed value of the corresponding measurement. Therefore, the weight of the term corresponding to the m -th measurement in equation (2.2) can be set as inverse of the square of standard deviation i.e. inverse of variance as in equation (2.10)

$$w_m = \frac{1}{\sigma_m^2} \quad (2.10)$$

The variance (σ^2) terms for all the measurements are usually assembled as the elements of a diagonal matrix $[R]$ known as measurement error covariance matrix as in equation (2.11).

$$[R] = \begin{bmatrix} \sigma_1^2 & & & \\ & \sigma_2^2 & & \\ & & \ddots & \\ & & & \sigma_m^2 \end{bmatrix} \quad (2.11)$$

Equations (2.10) and (2.11) can be combined as:

$$[R]^{-1} = \begin{bmatrix} w_1 & & & \\ & w_2 & & \\ & & \ddots & \\ & & & w_m \end{bmatrix} \quad (2.12)$$

In both the equations (2.11) and (2.12), the matrices are of dimension $M \times M$ i.e. $m = 1$ to $m = M$; the total number of measurements being M as mentioned in section 2.2.

2.2.2 Example of WLS State Estimation

In Figure 2.1 a 2-bus d.c. network with two nodal current measurements z_1 and z_2 and a branch current measurement z_3 has been shown.

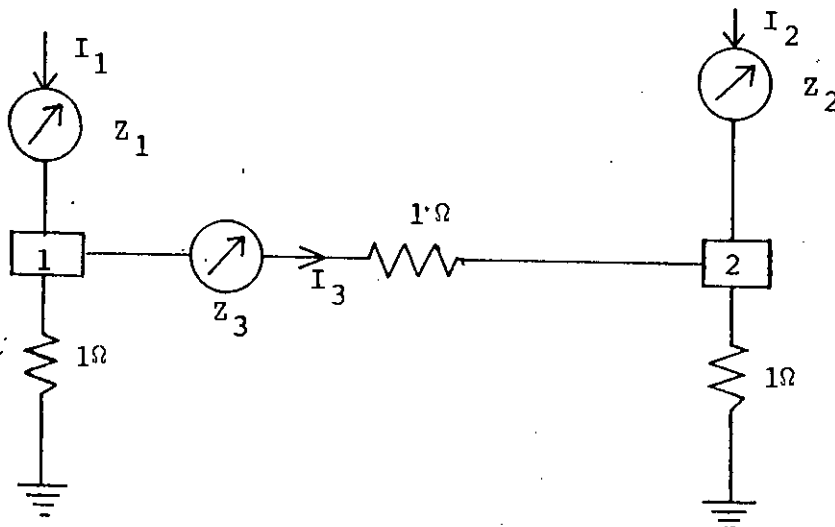


Figure 2.1: A 2-bus d.c. network.

The observed values of measurements have been assumed as

$$z_1 = 4 \text{ ampere}$$

$$z_2 = -1.5 \text{ ampere}$$

$$z_3 = 2.5 \text{ ampere}$$

The weight for each of the measurements has been assumed to be unity so that

$$[R]^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad (2.13)$$

It is required to estimate the nodal voltages.

In this case the state variables are

$$[X] = \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \quad (2.14)$$

while the equations relating the true values of measurements to the state variables are linear and can be derived applying kirchhoff's law as follows:

$$\begin{aligned} h_1 = I_1 &= 2v_1 - v_2 \\ h_2 = I_2 &= -v_1 + 2v_2 \\ h_3 = I_3 &= v_1 - v_2 \end{aligned} \quad (2.15)$$

Equations (2.15) can be stated in matrix form as in equation

(2.16)

$$[H] = \begin{bmatrix} 2 & -1 \\ -1 & 2 \\ 1 & -1 \end{bmatrix} \quad (2.16)$$

The observed measurement values can be expressed as

$$[Z] = \begin{bmatrix} 4 \\ -1.5 \\ 2.5 \end{bmatrix} \quad (2.17)$$

Now the gain matrix $[G]$ is

$$\begin{aligned} [G] &= [H]^T [R]^{-1} [H] \\ &= \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \end{bmatrix}^T \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ -1 & 2 \\ 1 & -1 \end{bmatrix} \\ &= \begin{bmatrix} 6 & -5 \\ -5 & 6 \end{bmatrix} \quad (2.18) \end{aligned}$$

Then applying equation (2.7)

$$\begin{aligned} [\hat{X}] &= [G]^{-1} [H]^T [R]^{-1} [Z] \\ &= \begin{bmatrix} 6 & -5 \\ -5 & 6 \end{bmatrix}^{-1} \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 4 \\ -1.5 \\ 2.5 \end{bmatrix} \\ &= \frac{1}{11} \begin{bmatrix} 6 & 5 \\ 5 & 6 \end{bmatrix} \begin{bmatrix} 2 & -1 & 1 \\ -1 & 2 & -1 \end{bmatrix} \begin{bmatrix} 4 \\ -1.5 \\ 2.5 \end{bmatrix} \end{aligned}$$

$$= \frac{1}{11} \begin{bmatrix} 7 & 4 & 1 \\ 4 & 7 & -1 \end{bmatrix} \begin{bmatrix} 4 \\ -1.5 \\ 2.5 \end{bmatrix}$$

$$\text{Or, } [\hat{X}] = \begin{bmatrix} 2.23 \\ 0.27 \end{bmatrix} \quad (2.19)$$

is the required estimate of nodal voltages.

2.3 Fast Decoupled WLS Algorithm¹²

The basic WLS algorithm defined by Equation (2.3) has the following main disadvantages.

- a) A large amount of computer storage is required because the gain matrix $[G]$ has a size $2n \times 2n$ when n is the number of nodes such that $N = 2n$; N being the number of state variables.
- b) The computing time is excessive due to evaluation of the gain matrix using the state vector dependent Jacobia matrix $[H]$ and its inversion in each iteration.

To overcome the above mentioned difficulties a grouping of the measurements into active $[Z_p]$ and reactive $[Z_q]$ sets [equations (2.20) and (2.21)] is made so as to express the Jacobian matrix $[H]$ of equation (2.3) in partitioned form as in equation (2.22).

$$[Z_p] = [h_p(\theta, v)] + [\xi_p] \quad (2.20)$$

$$[Z_q] = [h_q(\theta, v)] + [\xi_q] \quad (2.21)$$

$$[H(\theta, v)] = \begin{bmatrix} [H_{pp}(\theta, v)] & [H_{pq}(\theta, v)] \\ [H_{qp}(\theta, v)] & [H_{qq}(\theta, v)] \end{bmatrix} \quad (2.22)$$

$$[H_{pp}(\theta, v)] = \frac{\partial h_p}{\partial \theta}, \quad [H_{pq}(\theta, v)] = \frac{\partial h_p}{\partial v} \quad (2.23)$$

$$[H_{qp}(\theta, v)] = \frac{\partial h_q}{\partial \theta}, \quad [H_{qq}(\theta, v)] = \frac{\partial h_q}{\partial v} \quad (2.24)$$

In the Jacobian matrix of equation (2.22) the off-diagonal blocks i.e. submatrices $[H_{pq}]$ and $[H_{qp}]$ can be neglected on the basis of weak coupling which exists

- i) between change in phase angles and that in reactive powers
- and ii) between change in voltage magnitudes and that in active powers.

Using the decoupled Jacobian

$$\text{matrix} \begin{bmatrix} [H_{pp}(\theta, v)] & \\ & [H_{qq}(\theta, v)] \end{bmatrix}$$

and two measurement sets ($[Z_p]$, $[Z_q]$) as in equations (2.20) and (2.21), the equation (2.3) can be expressed as:

$$[A_{pp}(\hat{\theta}^i, \hat{v}^i)] [\Delta \hat{\theta}]^{i+1} = [H_{pp}(\hat{\theta}^i, \hat{v}^i)]^T [R_{pp}]^{-1} [\Delta Z_p]^{i+1} \quad (2.25)$$

$$[A_{qq}(\hat{\theta}^i, \hat{v}^i)] [\Delta \hat{v}]^{i+1} = [H_{qq}(\hat{\theta}^i, \hat{v}^i)]^T [R_{qq}]^{-1} [\Delta Z_q]^{i+1} \quad (2.26)$$

In equations (2.25) and (2.26) the state vector $[X]$ has been expressed in terms of voltage magnitude vector $[V]$ and phase angle vector $[\theta]$. Also,

$$[A_{pp}(\hat{\theta}^i, \hat{v}^i)] = [H_{pp}(\hat{\theta}^i, \hat{v}^i)]^T [R_{pp}]^{-1} [H_{pp}(\hat{\theta}^i, \hat{v}^i)] \quad (2.27)$$

$$[A_{qq}(\hat{\theta}^i, \hat{v}^i)] = [H_{qq}(\hat{\theta}^i, \hat{v}^i)]^T [R_{qq}]^{-1} [H_{qq}(\hat{\theta}^i, \hat{v}^i)] \quad (2.28)$$

$$[\Delta z_p]^{i+1} = [z_p] - [h_p(\hat{\theta}^i, \hat{v}^i)] \quad (2.29)$$

$$[\Delta z_q]^{i+1} = [z_q] - [h_q(\hat{\theta}^i, \hat{v}^i)] \quad (2.30)$$

$$[\Delta \hat{\theta}]^{i+1} = [\hat{\theta}]^{i+1} - [\hat{\theta}]^i \quad (2.31)$$

$$[\Delta \hat{V}]^{i+1} = [\hat{V}]^{i+1} - [\hat{V}]^i \quad (2.32)$$

It has been shown in section A.3 of Appendix A that based on a number of simplifications known as fast decoupling techniques, equation (2.27) and (2.28) can be refined to provide an algorithm with state vector independent Jacobian matrices. This algorithm is termed fast decoupled WLS Algorithm and expressed as in the equations (2.33) and (2.34)

$$[H_{pp}]^T [R_{pp}]^{-1} [H_{pp}] ([\hat{\theta}]^{i+1} - [\hat{\theta}]^i) = [H_{pp}]^T [R_{pp}]^{-1} ([z_p] - [h_p(\hat{\theta}^i, \hat{v}^i)])$$

(2.33)

$$\begin{aligned}
& [H'_{qq}]^T [R_{qq}]^{-1} [H'_{qq}] ([\hat{V}]^{i+1} - [\hat{V}]^i) \\
& = [H'_{qq}]^T [R_{qq}]^{-1} ([Z'_q] - [h'_q(\hat{\theta}^{i+1}, \hat{V}^i)]) \quad (2.34)
\end{aligned}$$

where the Jacobian and measurement error covariance matrices are with constant elements.

2.3.1 Computational Aspects of Fast Decoupled Algorithm

a) The active Jacobian Matrix $[H_{pp}]$ and the reactive Jacobian matrix $[H'_{qq}]$ are independent of state variables and evaluated before the start of the iterative process. Hence the corresponding gain matrices are also constant and inverted only once. Hence the time per iteration and overall computing time of this method are greatly reduced compared to those of the basic WLS algorithm.

b) In an iteration (i+1) firstly the vector $[\hat{\theta}]$ is updated using active measurement residuals $[\Delta Z'_p]$ evaluated at both $[\hat{\theta}]$ and $[\hat{V}]$ of previous iteration i. Then $[\hat{V}]$ is computed in the same iteration (i+1) using the reactive measurement residuals $[\Delta Z'_q]$ evaluated at $[\hat{V}]$ of previous iteration i and updated $[\hat{\theta}]$ of current iteration (i+1).

c) The algorithm stops the iterative process when all the elements of both $[\Delta \hat{\theta}]^{i+1} = [\hat{\theta}]^{i+1} - [\hat{\theta}]^i$ and $[\Delta \hat{V}]^{i+1} = [\hat{V}]^{i+1} - [\hat{V}]^i$ are less than or equal to the tolerance margin. This

may happen either at the end of a number of full iterations or a number of full iterations plus a half which does not require updating of $\{\hat{V}\}$.

d) The storage requirement is largely reduced as only two diagonal blocks of the gain matrix $[G]$ and two submatrices of the Jacobin Matrix $[H]$ are to be stored.

2.4 Bad Data Suppression

In this technique^{11,14} each of the computed measurement residuals is checked against a carefully chosen threshold (break-point) value λ . If the residual exceeds the threshold the corresponding measurement is a suspect one and is assigned less weight else the same weight as in the WLS criterion. This process of reweighting can be confined to only the right hand side of the equation (2.3) which defines a basic WLS algorithm. As a result the general expression for a WLS estimator transformed into a Bad Data Suppression (BDS) algorithm can be written as:

$$[\hat{X}]^{i+1} - [\hat{X}]^i = [G]^{-1} [H]^T [R]^{-1} [D]^{i+1} [\rho]^{i+1} \quad (2.35)$$

where the notations have their usual significance excepting

$[D]$ - a diagonal matrix with elements (d_m) as a function of corresponding measurement residual $r_m = z_m - h_m(\hat{x}^i)$

$[\rho]$ - a vector with elements (ρ_m) as function of r_m

$m = 1, 2, \dots, M$ denotes measurement serial

A fast decoupled BDS algorithm results when equation (2.35) is extended to equations (2.33) and (2.34) representing the fast decoupled WLS algorithm.

The quadratic square root criterion¹⁴ used for reweighting expresses the elements d_m in the matrix [D] and ρ_m in the vector [ρ] by equations (2.36) and (2.37).

$$\text{If } \left| \frac{r_m}{\sigma_m} \right| > \lambda \text{ then}$$

$$d_m = \left(4 \left| \frac{r_m}{\lambda \sigma_m} \right| - 3 \right)^{-1/2} \left| \frac{r_m}{\lambda \sigma_m} \right|^{-1} \quad (2.36)$$

$$\rho_m = \pm \lambda \sigma_m \left(4 \left| \frac{r_m}{\lambda \sigma_m} \right| - 3 \right)^{1/2}$$

else

$$d_m = 1$$

$$\rho_m = r_m \quad (2.37)$$

In equation (2.36) positive sign is used for ρ_m when r_m is positive while negative when r_m is negative.

The choice of the threshold λ determines the size of suspect bad data to be suppressed. Since the usual measurement noise obeying Gaussian distribution law are within $\pm 3 \sigma_m$ range¹⁸ for a probability of 99.7%, the bad data with errors outside, this range will be suppressed for a value chosen for λ not much greater than 3.0. The higher the value of λ the less is the suppression effect.

2.5 Observability¹⁹

The observability of a system is defined as the possibility of determining its state variables using the available measurements. The concept of observability can be understood by explaining the relation between metering configuration and the network configuration.

Let k , l , m and r be four nodes with r chosen as the reference node as in Figure 2.2.

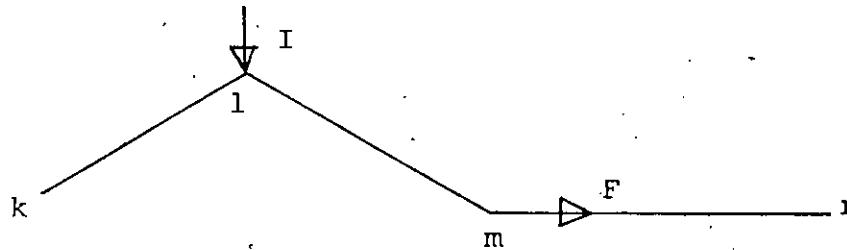


Figure 2.2: Demonstration of relation between measurement locations and network configuration

Let an injection measurement I and a flow measurement F be respectively placed at nodes l and m . Then the measurement I at node l provides a direct or an indirect path (connection) between any two nodes of the group k , l , m . The flow measurement F at node m provides a path between m and the reference node r . Since m is already measurement-connected to nodes k and l it can be said that each of the nodes k , l and m has a path to the reference node either directly or indirectly.

Similarly if all the nodes of a system under consideration are measurement-connected to the reference directly or indirectly then in the Jacobian matrix $[H]$ none of the columns, each of which corresponds to a node with unknown state variable, will have completely zero elements. As a result the inverse of the gain matrix $[G] = [H]^T [R]^{-1} [H]$ will exist and the equation (2.3) defining the basic WLS estimator can also be solved.

For a fast decoupled WLS estimator the observability is determined by considering the active set (active injections and line flows) and the reactive set (reactive injections, reactive flows and voltage magnitudes) of measurements separately corresponding to two gain matrices: active and reactive. If any one set fails the test for observability, the system as a whole is unobservable.

2.6 Conclusions

The basic weighted least square algorithm though provides the most accurate estimates, takes excessive computing time and storage. The fast decoupled WLS algorithm has been reported to take on average a few iterations more than the basic WLS algorithm but produces results with an accuracy close to that of basic WLS as well as adequate for practical purpose involving less computer storage and taking a less overall computing time due to a reduced time per iteration.

The usual random measurement noise obeying Gaussian

probability distribution is inherently filtered by a WLS estimator due to the processing of a redundant measurement set and use of weighting factor equal to the inverse of the square of the standard deviation of the corresponding measurement noise. But the gross measurement errors outside the Gaussian distribution require to be suppressed incorporating a reweighting criterion known as quadratic square root criterion with a WLS estimator.

CHAPTER 3
DEVELOPMENT OF COMPUTER PROGRAM

3.1 Introduction

The main computations involved in a fast decoupled BDS algorithm can be classified as fixed computations, iterative calculations together with bad data suppression and output of the estimation results.

In an on-line environment, when a state estimator is implemented on the computer of a power system control (dispatch) centre, the measurements are directly acquired from the system and fed as input to the state estimation program. The state estimator is also interfaced with two more on-line softwares namely network configurator and the data base. The former provides information on the current configuration of system network while the latter provides network parameters and information on meter placements i.e. locations of the acquired measurements. Also the output of the state estimator is stored in the data base for accession by other application programs.

Before on-line implementation of a software for power system control, it is a common practice to test the developed prototype using simulated data. Simulation test of the computer code developed for a state estimation algorithm is done adopting Monte Carlo approach¹⁶. A metering pattern comprising active and reactive nodal power injections, line power flows and voltage measurements at various nodes is selected. The 'true' values of the measurement quantities are obtained from a load flow solution of the test system. Also the values

of network parameters used by load flow program are accessed by the state estimation program through common blocks. The standard deviation for each of the measurements is provided to a Gaussian random number generation subroutine¹⁶ to obtain a quantity which represents a zero mean random error. This quantity is added with the 'true' value of the corresponding measurement to simulate telemetered raw data. Some of the measurement values are made largely erroneous to simulate bad data. Then this simulated set of 'telemetered' and 'erroneous' measurements is provided to the developed state estimator as input. In this way the selected metering pattern is tested many times, for instance, 30, using different sets of errors to corrupt 'true' values of measurements in each simulation.

3.2 Computer Program for Simulation Test of the Fast Decoupled State Estimator

The basic scheme for interfacing various groups of subroutines written for Monte Carlo simulation test of the fast decoupled BDS algorithm has been presented in the form of a flow chart in Figure 3.1. The main program and all the subroutines were written in standard FORTRAN 77 language.

3.2.1 Computation of Measurement Weighting Factors

Use of standard deviations is required not only in simulating measurement errors but also in computing the weights

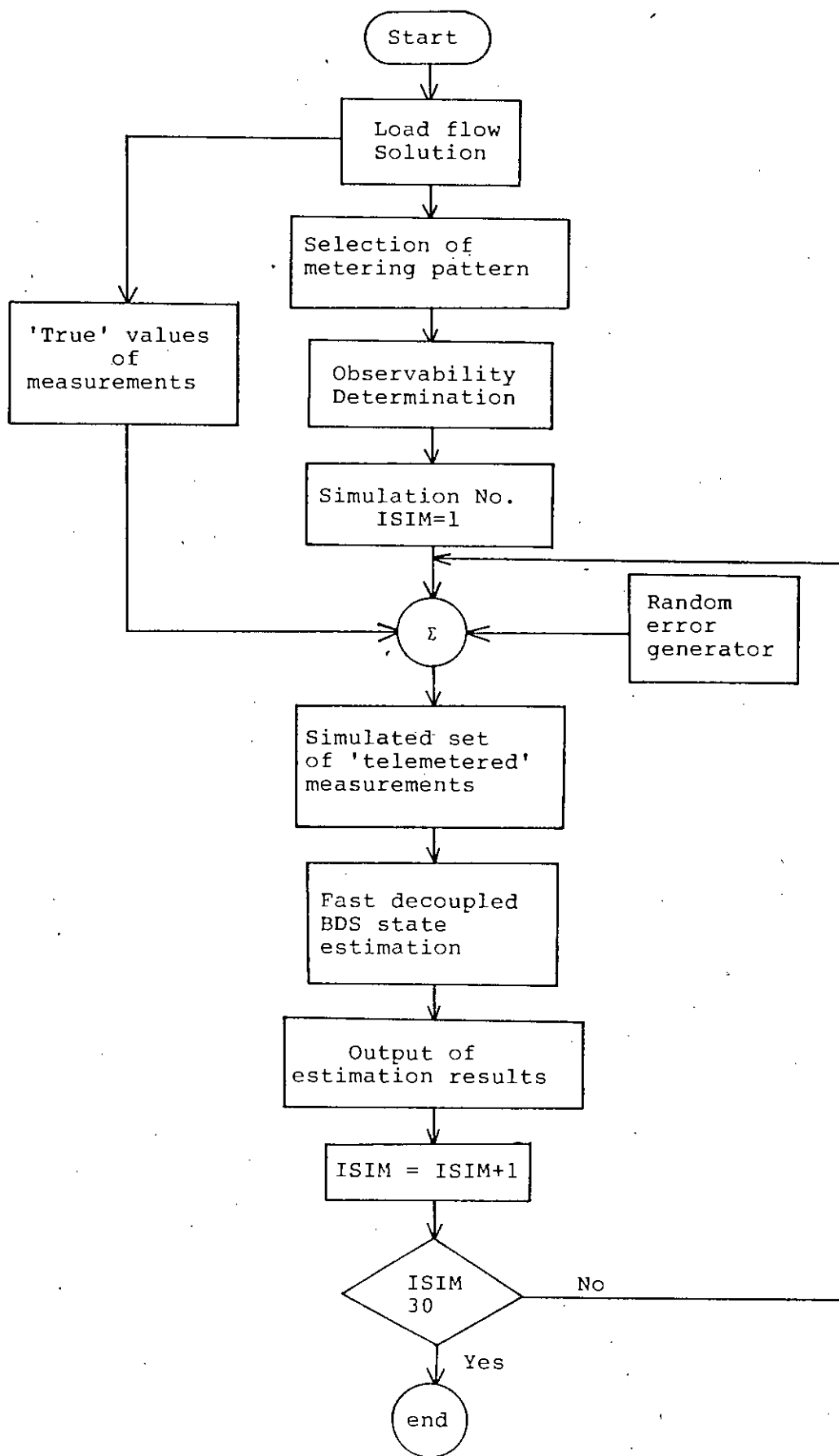


Figure 3.1: The basic scheme for Monte Carlo simulation test of the fast decoupled BDS estimator.

for the measurements of the selected metering pattern. The numerical values of σ can be calculated from the knowledge of the accuracy of the instrumentation and the telemetry system. The accuracies are presented by the manufacturers as the maximum expected errors. Under a Gaussian distribution the maximum error will fall within a range of $\pm 3 \sigma$ (3 standard deviations) for a probability of 99.7%. Therefore, using typical accuracies the standard deviation σ_m for the m -th measurement, if it is a line flow measurement, can be calculated²⁰ as

$$\sigma_m = 1/3 (a_1 |S| + a_2 D + a_3 D) \quad (3.1)$$

where,

S - complex quantity (MVA) representing 'true' flow value

D - full scale (MVA) deflection corresponding to rated transducer output

$a_1 = 2\%$ to 12% due to current and potential transformer errors

$a_2 = 0.25\%$ to 0.35% due to transducer error

$a_3 = 0.1\%$ due to A/D converters

σ_m = in units of MVA

In the present work it has been considered that

$$a_1 = 3\%$$

$$a_2 = 0.25\%$$

$$a_3 = 0.1\%$$

$$D = 1000 \text{ MVA}$$

So that equation (3.1) becomes

$$\sigma_m \simeq (0.01 |S| + 1.2) \quad (3.2)$$

where

σ_m - in units of MVA

If all the quantities in equation (3.2) are expressed in per unit (p.u.) with a base MVA equal to 'MVABASE' then σ_m can be computed in p.u. as in equation (3.3)

$$\sigma_m = 0.01 |S| + F \quad (3.3)$$

where

$$F = 1.2/MVABASE \quad (3.4)$$

σ_m and S are in p.u.

It is a practice to calculate the standard deviation (in p.u.) corresponding to an injection measurement as in equation (3.5)

$$\sigma_m = F \quad (3.5)$$

where F is as in equation (3.4)

For voltage measurements the standard deviation is

$$\sigma_m = 0.001v_m \quad (3.6)$$

where v_m is the 'true' magnitude of the corresponding voltage measurement in p.u.

It is worth mentioning that during on-line implementation as the measurements are not a simulated set rather directly acquired from the system the 'true' values are not available and instead typical observed (measured) values are used.

3.2.2 Observability Determination

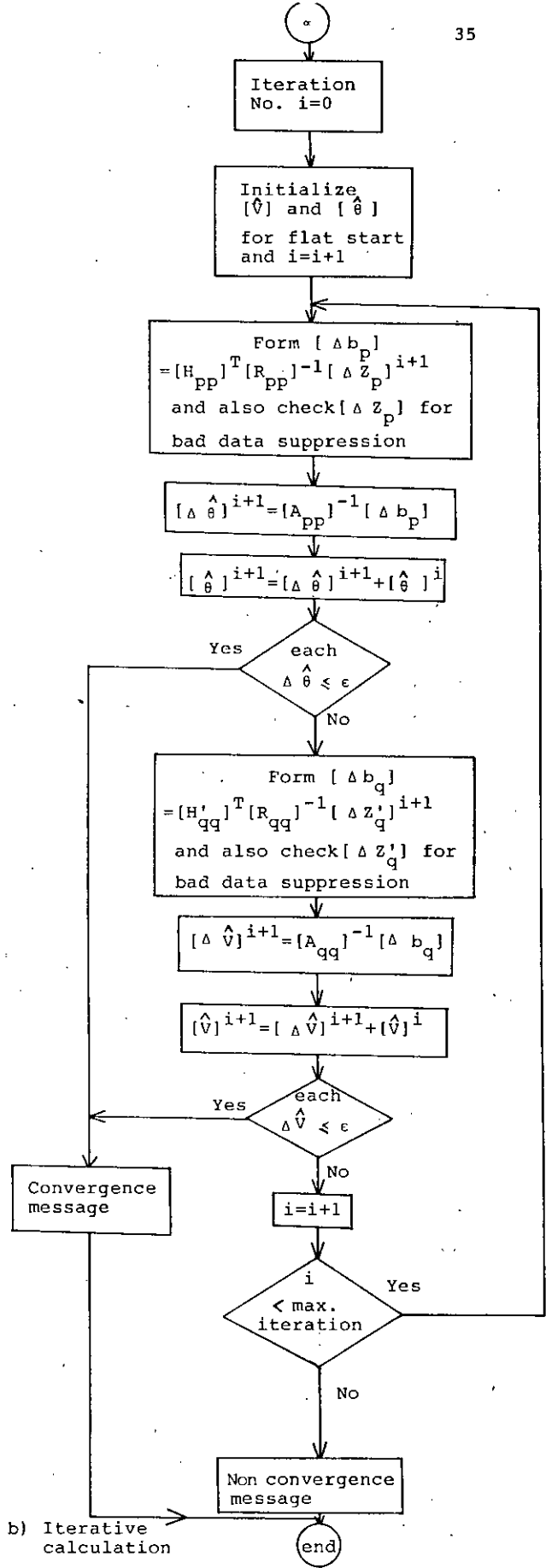
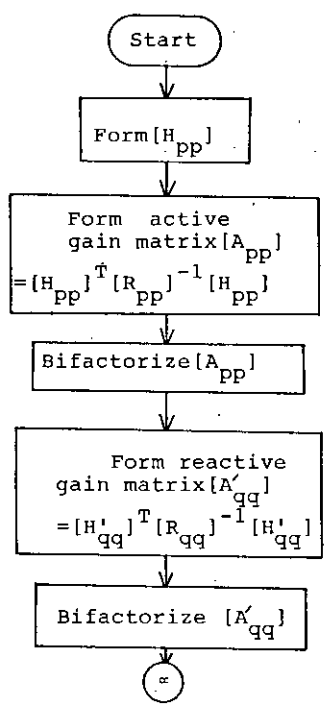
The test for observability has been done by assigning a code number to any one of the nodes. Then the group of nodes connected to this node directly or indirectly through measurements were identified and given the same code. In this way a check was made if all the nodes of a system attained a common code and the system became observable. The test has been conducted twice by considering the active and reactive sets of measurements separately for the fast decoupled estimator. It should be noted this method of observability test is a simple and quick one involving only logical statements and no floating point calculations.

3.3 Flow Chart for the Fast Decoupled BDS' State Estimation

Various steps of the fast decoupled BDS state estimation algorithm have been programmed by the present research work according to the flow chart presented in Figure 3.2.

3.3.1 Fixed Calculations

Fixed calculations refer to those computations which need to be accomplished only once and remain valid as long as the measurement configuration and network topology remain unchanged.



a) Fixed calculation

b) Iterative calculation

Figure 3.2: Flow chart for fast Decoupled BDS State Estimation Algorithm.

These are mainly formation of active and reactive Jacobian matrices respectively $[H_{pp}]$ and $[H'_{qq}]$, and corresponding gain matrices as in equations (2.33) and (2.34). All these matrices have constant parameter elements and are predominantly sparse i.e. the greater proportion of the elements are zero. Therefore, in the present research work, storage of Jacobian matrices and inversion of the gain matrices have been done using sparsity oriented programming.

In the sparsity¹⁵ exploited scheme only the nonzero elements are recorded using row and column index arrays instead of conventional two dimensional matrices. Moreover instead of conventional inversion technique a more efficient method called sparsity directed bifactorization¹⁵ is used. This saves both computer storage and time.

3.3.2 Iterative Calculations

Iterative calculations refer to the computations required to be made in each iteration. These are mainly formation of active and reactive measurement residual vectors $[\Delta Z_p]$ and $[\Delta Z'_q]$, check of residuals for bad data suppression, updating state variables ($[\hat{\theta}]$ and $[\hat{V}]$) and convergence test.

3.4 Conclusions

The computer code developed for the fast decoupled BDS algorithm in the present research project is by nature a program to make use of simulated data. During on-line implementation only the program for the state estimator together with that for observability is to be adapted for interfacing with the data base and network configurator on a power system control computer.

Applications of sparsity exploitation techniques in the developed program has enhanced its efficiency in terms of computer storage and time.

CHAPTER 4

PRESENTATION OF RESULTS

4.1 Introduction

The computer program developed by the present research work for simulation test of the fast decoupled BDS state estimation algorithm has been applied on the BPDB grid system (as of June 1989) using the mainframe computer IBM 4331/KO2 at BUET.

The simulation test has been done extensively to evaluate various performance such as convergence characteristics, accuracy of estimates and bad data processing of the algorithm on BPDB system corresponding to a number of metering configurations. The test system consisting of two interconnected zones viz: east and west, comprises a total of 81 nodes (buses) and 114 lines. The metering pattern was varied by choosing locations of injection (difference between generation and load at a bus), flow and voltage magnitude measurements at various nodes. The estimation has been performed only after the test system was proved observable with respect to the selected measurement configuration. Each metering pattern was tested 30 times using different sets of random errors in the way explained in sections 3.1 and 3.2 of Chapter 3.

The results of the comprehensive simulation study made by the present work have been presented in the following sections in the form of general comments together with typical selected numerical values to illustrate the average performance of the developed estimator.

The test system data and some sample results have been shown respectively in Appendix B and Appendix C.

4.2 Measurement Configurations

Some typical metering patterns used in the simulation test are shown in the Table 4.1.

Table 4.1: Typical measurement configurations.

Measurement Configuration No.	Location of measurements			Total No. of measurements M	Redundancy ratio $\eta = \frac{M}{N}$
	Injection	Flow	Voltage		
1	at B2	at SE	at GB	287	1.79
2	at GB	at SE	at GB	303	1.89
3	at GB and SB	at SE	at GB and SB	321	2.01
4	-	at SE and RE	-	456	2.85
5	-	at SE and RE	at GB	481	3.01

The significance of various symbols used in Table 4.1 are as follows:

- B2 : each of the buses having 2 lines connected to it;
there are 17 such buses in the BPDB system.
- GB : each of the buses having generator connected to it;
there are 25 such buses in the BPDB system.
- SB : each of the buses having no generator or load connected to it; these are termed as switching bus and there are 6 such buses in the BPDB system.
- SE : sending end of each line.
- RE : receiving end of each line.
- N : number of state variables for 81-bus BPDB system is 160; this includes the voltage magnitude and phase angle of each node excepting the one which has been treated as the reference node with a voltage magnitude assumed to be known and a zero phase angle.

Each of the injection and flow measurements was considered in the form of both active (MW) and reactive (MVAR) components.

4.3 Convergence Characteristics

The fast decoupled BDS algorithm was tested using a tolerance margin of 10^{-6} between the estimates from two successive iterations and a flat start $v/\theta = 1.0/0$ in the first iteration. Table 4.2 provides some typical results on the convergence property of the algorithm in terms of number of iterations when all the measurements were corrupted with random errors.

Table 4.2: Results on convergence of the fast decoupled BDS estimator.

Metering pattern No.	Redundancy ratio	No. of iterations
1	1.79	10.0
2	1.89	9.0
3	2.01	9.5
4	2.85	10.0
5	3.01	8.5

In table 4.2, the number of iterations in some cases are a whole number plus a half (e.g. 8.5) to imply that the estimation for voltage magnitudes was not required in the last iterations.

In general, simulation tests of the selected metering

patterns have shown that the algorithm took on average 8 to 10 iterations depending upon the redundancy ratio, amount of errors and locations of measurements. Although in some cases metering patterns having a higher redundancy ratio converged in less number of iterations in other cases metering patterns with a lower redundancy ratio but with the measurements placed uniformly around each node, have shown better convergence.

4.4 Accuracy of Estimates

The accuracy and quality of estimates provided by the test algorithm is judged²⁰ by computing the post-estimation value of WLS objective function $J(\hat{x})$ as follows:

$$J(\hat{x}) = \frac{1}{M} \sum_{m=1}^M w_m (z_m - h_m(\hat{x}))^2 \quad (4.1)$$

where

M - number of measurements

z_m - observed value of the m -th measurement

w_m - weighting factor for the m -th measurement

$h_m(\hat{x})$ - value of the m -th measurement computed from the corresponding analytical expression using obtained estimates of the state variables.

In simulation test z_m is the value specified for the m -th measurement after corrupting it with error.

If the computed value of $J(\hat{x})$ is less than the corresponding threshold value than the results of estimates are accurate and valid. If all the measurement errors follow the Gaussian distribution then $J(\hat{x})$ will have a chi-square distribution¹⁷ with $(M-N)$ degrees of freedom. Since the Gaussian measurement errors are within ± 3 standard deviations for a probability of 99.7% the corresponding threshold value²⁰ of $J(\hat{x})$ is

$$J(\hat{x})_{\text{threshold}} = \frac{(M-N) + 3 \sqrt{2(M-N)}}{M} \quad (4.2)$$

Table 4.3 shows some typical average values of $J(\hat{x})$ obtained in simulation tests conducted with only random noise corrupted measurements.

Table 4.3: Computed post-estimation vs. threshold values of $J(\hat{x})$ in presence of random measurement noise.

Metering pattern No.	Degrees of freedom M-N (N=160)	$J(\hat{x})$	
		computed value in p.u.	Threshold value in p.u.
1	127	0.3427	0.6091
2	143	0.4535	0.6393
3	161	0.4922	0.6692
4	296	0.6384	0.8091
5	321	0.6929	0.8253

The computed values of $J(\hat{x})$ were found to be less than the respective thresholds implying the acceptability of estimation results in terms of accuracy and quality.

At simulation stage few more performance indices²⁰ as shown by equations (4.3) through (4.5) can also be used to assess how close to the 'true' values are the estimates.

$$J_t(x) = \frac{1}{M} \sum_{m=1}^M w_m ((z_m)_{\text{true}} - h_m(\hat{x}))^2 \quad (4.3)$$

$$\mu_v = \frac{1}{N} \sum_{i=1}^N \left| (v_i)_{\text{true}} - \hat{v}_i \right| \quad (4.4)$$

$$\mu_\theta = \frac{1}{N} \sum_{i=1}^N \left| (\theta_i)_{\text{true}} - \hat{\theta}_i \right| \quad (4.5)$$

Table 4.4 shows the post-estimation and threshold values of the index $J_t(x)$ for the typical measurement configurations used in the present work.

The estimation results were close to the 'true' values and acceptable as evident from the computed values of the index $J_t(x)$ which were less than the respective thresholds.

The threshold values of $J_t(x)$ shown in Table 4.4 were

Table 4.4: Typical computed vs. threshold values of $J_t(x)$ in presence of random measurement noise.

Metering pattern No.	$J_t(x)$	
	Computed Value in p.u.	Threshold value in p.u.
1	0.6321	0.7444
2	0.4459	0.7051
3	0.4064	0.6656
4	0.3146	0.4685
5	0.2920	0.4442

obtained²⁰ using equation (4.6).

$$[J_t(x)]_{\text{threshold}} = \frac{N+3\sqrt{2N}}{M} \quad (4.6)$$

N was 160 for the BPDB system.

Table 4.5 includes the typical values obtained for the other two indices μ_v and μ_θ

Table 4.5 : Typical values of performance indices μ_v and μ_θ in presence of random measurement noise.

Metering pattern No.	μ_v in p.u.	μ_θ in radian
1	0.0015	0.0028
2	0.0010	0.0017
3	0.0009	0.0019
4	0.0017	0.0016
5	0.0011	0.0018

Very low values obtained for the performance indices μ_v and μ_θ imply that the estimates of state variables v/θ were close to the 'true' values.

The closeness of the estimates to the 'true' values can also be assessed by a comparison of the estimated quantities directly with their 'true' values.

Table 4.6 shows sample results on comparison of the estimates of voltages and phase angles at some of the nodes of BPDB grid system. Also typical values of estimates of some of the measurements have been shown in Table 4.7. The sample results shown in both the Tables were obtained for the metering pattern No. 1 in presence of random measurement noise.

The results shown in Table 4.6 and Table 4.7 confirm the closeness of estimated values to 'true' values and also an improvement over the specified (measured) values.

Table 4.6: Typical values of estimated voltages and phase angles at some of the nodes of the BPDB system for metering pattern No. 1

Node name	'True' voltage (p.u)	Estimated voltage (p.u.)	Est.-'True' voltage (p.u.)	'True' phase angle (degrees)	Estimated phase angle (degrees)	Est.-'True' phase angle (degrees)
KPTI2X40	1.0299	1.0290	-0.0009	3.0545	3.1432	0.0887
KHLN1X60	1.0299	1.0304	0.0005	-37.2741	-37.1378	0.1363
KAPTAL13	1.0020	1.0015	-0.0005	-1.7925	-1.5705	0.2220
SHJIBZ13	1.0170	1.0179	0.0009	-0.0804	-0.0859	-0.0055
GHRASL23	1.0252	1.0258	0.0006	-4.0673	-3.9397	0.1276

Table 4.7: Typical values of estimates of some of the measurements at selected nodes of the BPDB system for metering pattern No. 1

Node name	Type of measurement	'True' value	Specified value	Estimate value	Sp. - Est. values	Est.-True value
CHNRO13	Active injection	-18.89 (MW)	-17.95 (MW)	-18.43 (MW)	0.48 (MW)	0.46 (MW)
FRIDPR13	Reactive flow	11.85 (MVAR)	7.89 (MVAR)	11.11 (MVAR)	-3.22 (MVAR)	-0.74 (MVAR)
ULLONL13	Active flow	-40.00 (MW)	-40.64 (MW)	-40.12 (MW)	-0.52 (MW)	-0.12 (MW)
GHRASL13	Active flow	-63.76 (MW)	-61.48 (MW)	-63.15 (MW)	1.67 (MW)	0.61 (MW)

4.4.1 Accuracy vs. Redundancy

In general simulation tests have revealed that accuracy of estimates increased with the redundancy ratio of the measurement configuration. A quick assessment of accuracy vs. redundancy has been provided in Figures 4.1 and 4.2 by plotting the computed and threshold values of performance indices $J(\hat{x})$ and $J_t(x)$ respectively taken from Table 4.3 and Table 4.4.

It is evident from the Figures 4.1 and 4.2 that the computed value of performance index $J(\hat{x})$ increases with redundancy while that of $J_t(x)$ decreases. This is because of the fact that the former measures the difference between specified and estimated values of measurements while the latter assesses the difference between the 'true' and estimated values. As the redundancy increased the estimates became closer to the 'true' values and hence better than the specified i.e. erroneous values.

The increase in the threshold values of the index $J(\hat{x})$ with redundancy was consistent with the equation (4.2) which shows that higher the number of measurements higher will be the threshold. On the otherhand the threshold value of the index $J_t(x)$ which is inversely proportional to the number of measurements as in the equation (4.6), decreased with redundancy.

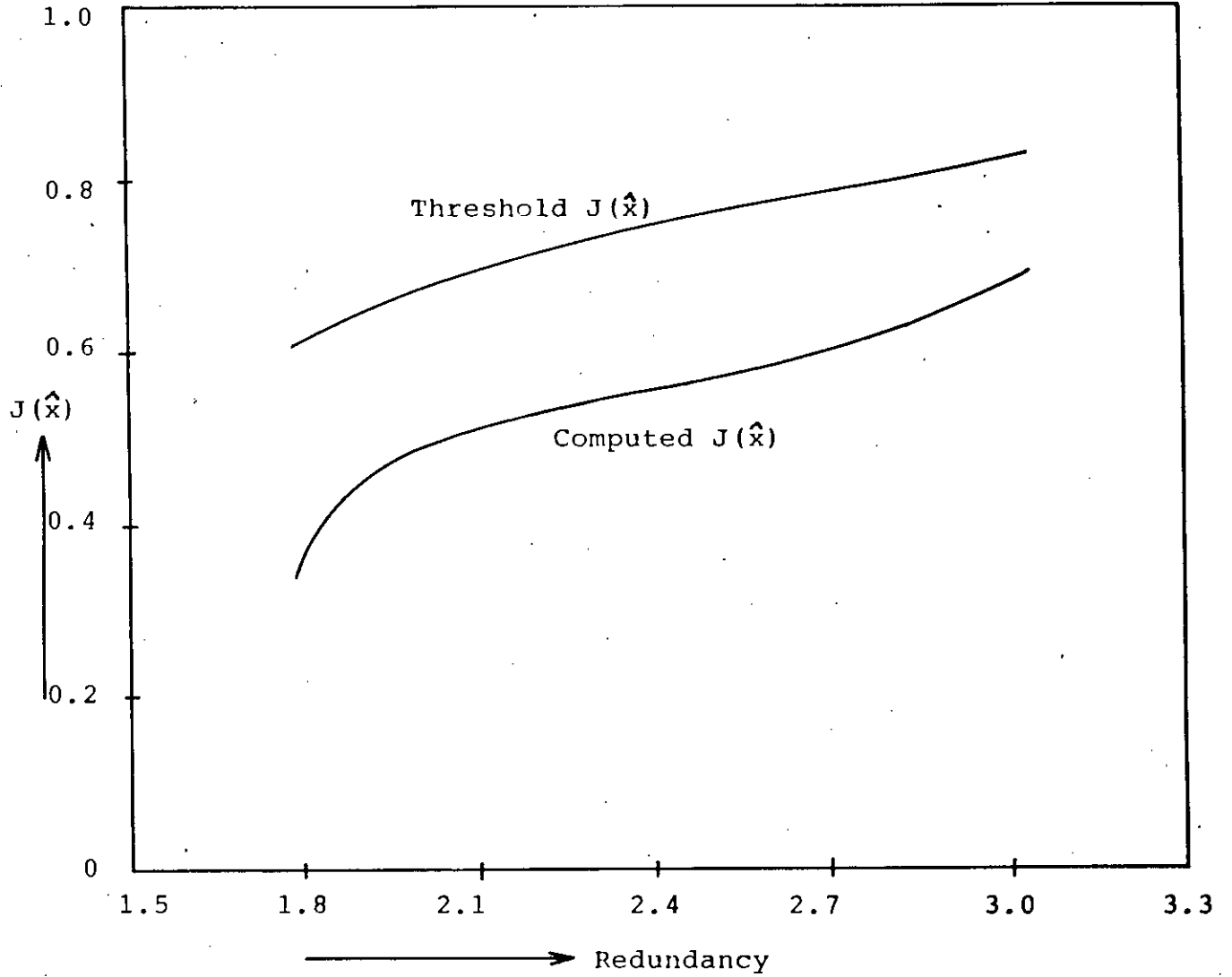


Figure 4.1: Variation of performance index $J(\hat{x})$ with redundancy.

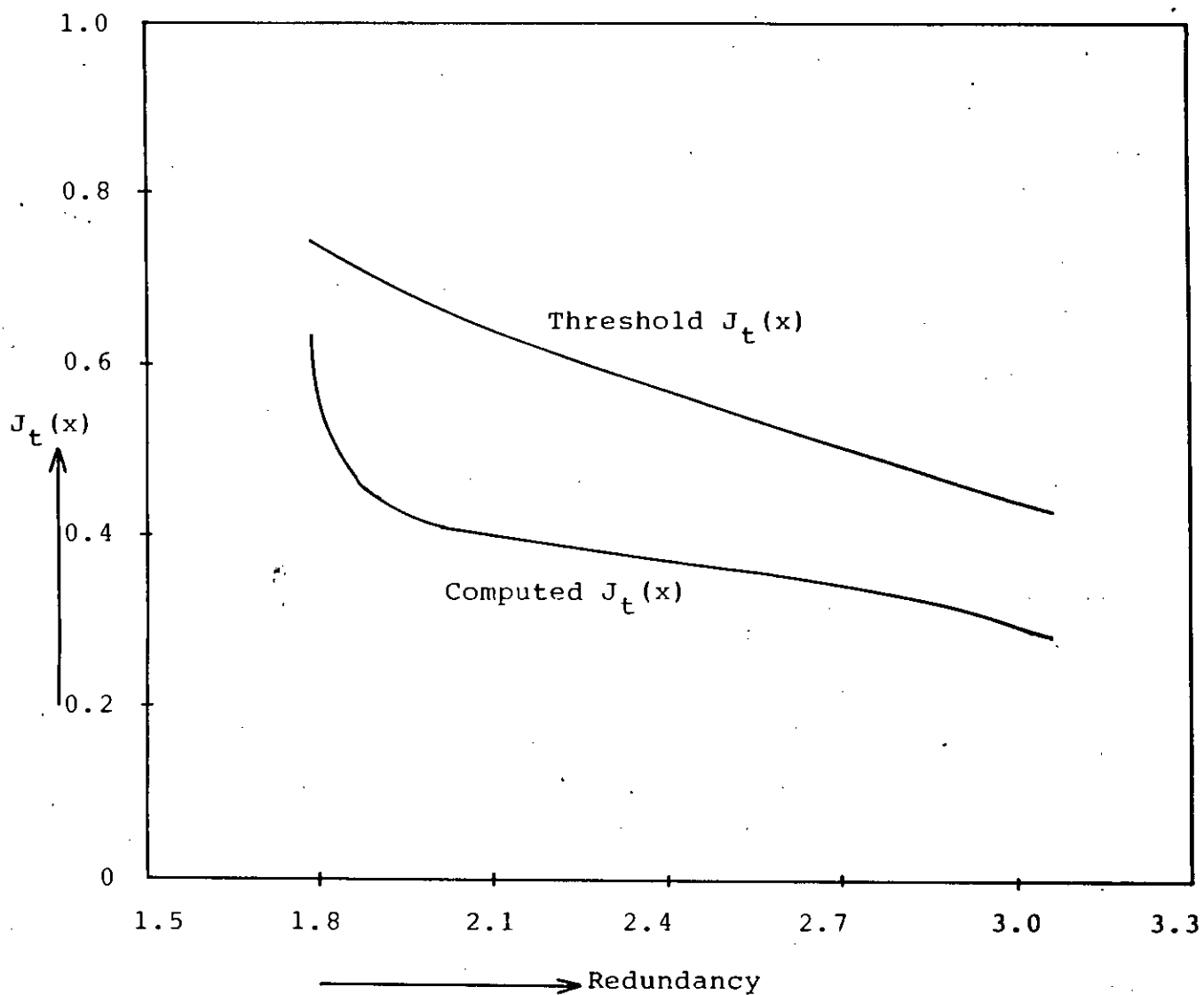


Figure 4.2: Variation of performance index $J_t(x)$ with redundancy.

4.5 Bad Data Performance

The algorithm was also tested by specifying zero values for some of the injection and flow measurements whose 'true' values were non zero. The rest of the measurements were random error corrupted.

Table 4.8 shows the performance of the algorithm for a typical case of bad data. The results were obtained with a value of 4.0 for the breakpoint explained in section 2.4 of Chapter 2. The tolerance margin used in the iterative process was 10^{-6} .

In general, the algorithm was able to suppress the bad data and produce estimates close to the 'true' values as evident from the value of the performance index $J_t(x)$ cited in Table 4.8. The computed value of the other index $J(\hat{x})$ was in excess of the threshold levels by a large margin, because the errors of some of the measurements were gross so that $J(\hat{x})$ did not have a chi-square distribution and also the specified values of those measurements (bad data) were widely different from the 'true' and hence estimated values.

The number of iterations required by the algorithm to converge was larger because of the presence of gross measurement errors in addition to random noise.

Table 4.8: Bad data performance for a typical case.

Metering pattern No.	Bad data Location	'True' Value	Specified Value	No.of itera- tion	Estimated Value	Performance indices in p.u.
5	An active and a reactive flow measurements at node CHNDRO13	-41.68 (MW)	0.0	29.0	-39.91 (MW)	$J(\hat{x})$ Computed: 2.2398 Threshold: 0.8253
		-22.37 (MVAR)	0.0		-22.95 (MVAR)	$J_t(x)$ Computed : 0.3302 Threshold: 0.4442

4.6 Computer Storage and Time

The approximate core memory requirements of the program developed in FORTRAN 77 for the fast decoupled BDS algorithm was worked out in terms of decimal words. The IBM 4331/K02 computer used for simulation tests in the present research has 32 bits word length.

The length of the program area was about 5500 words. The size of the data area was calculated considering the size of the test system i.e., 81 buses and 114 lines, and a maximum measurement redundancy ratio of 3.5. The integer type data area was approximately 8100 words and that for real type data was about 3240 words. It should be noted that the IBM 4331 computer has integer variable packing facility so that each word of 32 bit length can store more than one integer type data.

As regards to the computing time required for execution of the developed program it is worth mentioning that the program has been run from a time shared terminal system interfaced with the computer so that the CPU time recorded varied at different parts of the day and week depending upon job traffic. On average the total execution time of the state estimation program was about 10 seconds of which 3 seconds for fixed computations, 2 seconds for calculations in all the iterations and 5 seconds for writing the output on peripheral device. In general, it was observed that the total CPU time requirement of the program varied insignificantly with redundancy ratio.

4.7 Conclusions

The performance of the fast decoupled BDS state estimator on a practical system i.e. BPDB grid network has been evaluated by conducting Monte Carlo simulation tests extensively using various metering patterns with different redundancies.

The ability of the algorithm to converge with various sets of random and gross measurement errors has established its robustness. Use of a number of performance indices to assess the validity of the estimates has given positive results. The algorithm was successful in handling the bad data problem.

The variation of convergence, accuracy and computer time with redundancy has also been studied. Although CPU time varied insignificantly with redundancy the accuracy was found to increase with the same.

The convergence characteristics depended not only upon the redundancy but also on the uniformity of placing the measurements with respect to each node of the test system and magnitudes of measurement errors. The requirement of a small time per iteration made it allowable for the algorithm to converge in a comparatively large number of iterations in presence of bad data.

The computer time and storage requirement of the developed state estimator can be considered suitable for on-line implementation even on a mini or microcomputer with available memory

capacity. Had sparsity exploitation techniques not been applied, the developed program would have required more time and memory.

CHAPTER 5
GENERAL CONCLUSIONS

5.1 Conclusions

These days a computer based monitoring and control scheme is considered by the electric utilities as the only viable means to ensure an economic and secure power supply to their customers. The basic prerequisite to carry out any computer analysis and take correct decision is the provision of a reliable and consistent information on system state. To this end a set of redundant raw data (measurements) acquired from the system through telemetry is processed by an algorithm known as state estimator.

Bangladesh Power Development Board has just undertaken a project on the installation of a computer aided despatch centre for the on-line monitoring and control of the national grid system. Keeping in view the requirement of sophisticated software alongside the computer hardware, the present research work has attempted an investigation into various aspects involved in the initial stage of implementing a state estimation software responsible for performing a primary function in the on-line control. This investigation has covered the choice of an algorithm which would take computer storage and time acceptable from on-line standard point of view and identification of the metering patterns which would be suitable for the chosen algorithm in terms of convergence, accuracy of estimates and bad data (gross error) processing capability.

The Fast Decoupled Bad Data Suppression algorithm has been chosen and tested extensively on the BPDB grid system adopting Monte Carlo simulation approach. The algorithm has been programmed

exploiting sparsity so as to preserve its computational efficiency. The simulation tests performed with a number of selected metering configurations has established its robustness and ability to give accurate estimates in presence of both random and gross measurement errors.

5.2 Suggestions for Further Research

The computer code developed for the state estimation algorithm in the present research is by nature a program to make use of simulated data. Hence an investigation should be made into its adaptation for on-line operation which involves its interface with network configurator, data base and man-machine interface.

Also an investigation into the provision of some pre-estimation procedure for identifying some of the measurements corrupted by very large errors, can be considered. The identified measurements may be replaced by pseudo-measurements with values chosen from past knowledge. As a result the overall accuracy of estimates can be expected to improve and also the task of post-estimation bad data processing will be easier.

Investigations should also be made into identification of network modelling error due to inconsistency in the information on circuit breaker status.

The design of a single metering pattern which would be optimum

in terms of number, type and location of measurements, accuracy requirement and cost of the instrumentation system with respect to the fast decoupled BDS state estimator, can be an interesting point for further investigation.

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APPENDIX A
MATHEMATICAL EXPRESSIONS AND DERIVATIONS
FOR THE BASIC AND THE FAST DECOUPLED
WLS STATE ESTIMATORS

A.1 Nonlinear Functions of State Variables

The expressions $h(x)$ for a nodal injection and a line flow have been shown in Table A.1 in relation to an elementary network branch model and in terms of associated state variables together with the branch parameters.

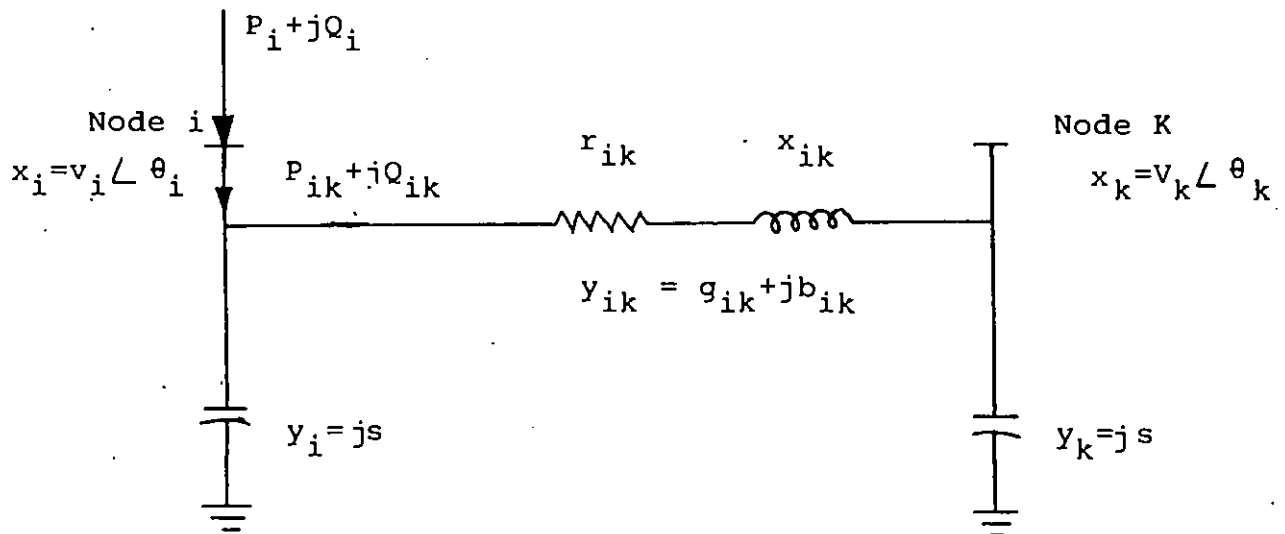


Table A.1: Expressions for a line flow and a nodal injection.

		$h(x)$
Line flow	P_{ik}	$P_{ik} = \text{Real part of } \{x_i (x_i y_i + (x_i - x_k) y_{ik})^*\}$ Or, $P_{ik} = v_i v_k [-g_{ik} \cos \theta_{ik} - b_{ik} \sin \theta_{ik}] + v_i^2 g_{ik}$
	Q_{ik}	$Q_{ik} = \text{Imaginary Part of } \{x_i (x_i y_i + (x_i - x_k) y_{ik})^*\}$ Or, $Q_{ik} = v_i v_k [-g_{ik} \sin \theta_{ik} + b_{ik} \cos \theta_{ik}] - v_i^2 (b_{ik} + s)$
Nodal injection		$P_i = \sum_{k \in \alpha_i} P_{ik}$ $Q_i = \sum_{k \in \alpha_i} Q_{ik}$

The meanings of some of the symbols used in Table A.1 are as follows. Other symbols carry their usual significance.

X_i	-	state variable of node i
Y_i	-	shunt susceptance of node i
Y_{ik}	-	series admittance of line i-k
r_{ik}	-	series resistance of line i-k
x_{ik}	-	series reactance of line i-k
g_{ik}	=	$\frac{r_{ik}}{r_{ik}^2 + x_{ik}^2}$ is series conductance of line i-k
b_{ik}	=	$-\frac{x_{ik}}{r_{ik}^2 + x_{ik}^2}$ is series susceptance of line i-k
j	-	imaginary operator
s	-	half of the charging susceptance in line i-k
θ_{ik}	-	difference between phase angles at nodes i and k : $\theta_i - \theta_k$
$k \in \alpha_i$	-	set of nodes k connected to node i by lines

A.2 Derivation of Equation (2.3)

Let the equation (2.2) be reconsidered as equation (A.1)

$$J(x) = \sum_{m=1}^M w_m (z_m - h_m(x))^2 \quad (\text{A.1})$$

Equation (A.1) can be rewritten in matrix form as:

$$J = ([Z] - [h(x)])^T [R]^{-1} ([Z] - [h(x)]) \quad (\text{A.2})$$

where the inverse of the matrix $[R]$ is diagonal with elements independent of state variables and equal to weighting factor w_m for m -th measurement as also stated in equation (2.12)

The nonlinear functions in vector $[h(x)]$ can be linearised by Taylor's series expansion about an initial estimate point $[x^0]$ with higher order terms neglected:

$$[h(x)] = [h(x^0)] + [H]([X] - [X^0]) \quad (\text{A.3})$$

where

$$[H] = \left. \frac{\partial h(x)}{\partial x} \right|_{x=x^0}$$

By subtracting both sides of equation (A.3) from the vector $[Z]$ it can be written as:

$$[Z] - [h(x)] = [Z] - [h(x^0)] - [H]([X] - [X^0]) \quad (\text{A.4})$$

Equation (A.4) can be written as:

$$[Z] - [h(x)] = [\Delta Z] - [H][\Delta X] \quad (\text{A.5})$$

where

$$[\Delta Z] = [Z] - [h(x^0)] \quad (\text{A.6})$$

$$[\Delta X] = [X] - [X^0] \quad (\text{A.7})$$

Substituting equation (A.4) for $([Z] - [h(x)])$ into equation (A.2) the WLS objective function J can be stated as:

$$J = ([\Delta Z] - [H][\Delta X])^T [R]^{-1} ([\Delta Z] - [H][\Delta X]) \quad (\text{A.8})$$

Differentiating the function J in equation (A.8) with respect to the state vector $[X]$ gives:

$$\frac{\partial J}{\partial X} = 2 \left(\frac{\partial}{\partial X} ([\Delta Z] - [H][\Delta X])^T \right) [R]^{-1} ([\Delta Z] - [H][\Delta X]) \quad (\text{A.9})$$

Taking transpose of both sides of equation (A.4) and then differentiating with respect to $[X]$ gives:

$$\frac{\partial [h(x)]^T}{\partial X} = [H]^T \quad (\text{A.10})$$

Taking transpose of both sides of equation (A.5) and then diffe-

differentiating with respect to $[X]$ gives:

$$-\frac{\partial [h(x)]^T}{\partial x} = \frac{\partial ([\Delta Z] - [H][\Delta X])^T}{\partial x} \quad (\text{A.11})$$

Combining equation (A.10) and (A.11) gives

$$\frac{\partial ([\Delta Z] - [H][\Delta X])^T}{\partial x} = -[H]^T \quad (\text{A.12})$$

Substituting equation (A.12) for $\frac{\partial ([\Delta Z] - [H][\Delta X])^T}{\partial x}$ into equation (A.9) gives:

$$\frac{\partial J}{\partial x} = -2[H]^T[R]^{-1}([\Delta Z] - [H][\Delta X]) \quad (\text{A.13})$$

Equating the derivative of objective function J shown in equation (A.13) to zero gives an expression for optimum estimate $[\hat{X}]$ of the state vector

$$[H]^T[R]^{-1}[H][\Delta \hat{X}] = [H]^T[R]^{-1}[\Delta Z] \quad (\text{A.14})$$

$$\text{Or, } [\Delta \hat{X}] = ([H]^T[R]^{-1}[H])^{-1}[H]^T[R]^{-1}[\Delta Z] \quad (\text{A.15})$$

$$\text{Or, } [\Delta \hat{X}] = [G]^{-1}[H]^T[R]^{-1}[\Delta Z] \quad (\text{A.16})$$

where

$$[\Delta \hat{X}] = [\hat{X}] - [X^0] \quad (\text{A.17})$$

$$[G] = [H]^T[R]^{-1}[H] \quad (\text{A.18})$$

It is to be noted that the estimate $[\hat{X}]$ has been obtained after linearizing the nonlinear functions $[h(x)]$ about an initial point $[X^0]$. Not necessarily will $[\hat{X}]$ be the desired estimate close to the 'true' value. Therefore the estimation process is to be repeated using $[\hat{X}]$ as the next point of linearisation to obtain a better solution. In this way the process defined by equation (A.16) can be iterated until each of the elements of $[\Delta X]$ is less than or equal to a prespecified small quantity termed tolerance margin. Equation (A.19) shows the general expression of the iterative algorithm.

$$[\Delta \hat{X}]^{i+1} = [G]^{-1} [H]^T [R]^{-1} [\Delta Z]^{i+1} \quad (\text{A.19})$$

$$\text{Or, } [\hat{X}]^{i+1} = [\hat{X}]^i + [G]^{-1} [H]^T [R]^{-1} [\Delta Z]^{i+1} \quad (\text{A.20})$$

where,

$$[H] = \left. \frac{\partial h}{\partial x} \right|_{x=\hat{x}^i} \quad (\text{A.21})$$

$$[G] = [H]^T [R]^{-1} [H] \quad (\text{A.22})$$

$$[\Delta Z]^{i+1} = [Z] - [h(\hat{x}^i)] \quad (\text{A.23})$$

Equation (A.20) is the required expression of the basic WLS state estimation algorithm defined by equation (2.3).

A.3 Fast Decoupling Techniques

Equations (A.24) to (A.30) shows the simplifications known as fast decoupling techniques.

- 1) The reactive set of measurement equations are divided by the voltage magnitude v_i at the bus i at which the corresponding measurement was taken. This results in the disappearance of the dependence of the related Jacobian terms on voltage magnitudes.

$$[z'_q] = [h'_q(\theta, v)] + [\xi'_q] \quad (\text{A.24})$$

where,

$$(z'_q)_m = (z_q/v_i)_m$$

$$(h'_q)_m = (h_q/v_i)_m \quad (\text{A.25})$$

$$(\xi'_q)_m = (\xi_q/v_i)_m$$

$m=1,2,\dots$number of reactive measurements.

- 2) Due to reformulation as in equation (A.24) the weighting matrix of reactive measurement residuals $[R_{qq}]^{-1}$ changes to $[R'_{qq}]^{-1}$ such that:

$$[R'_{qq}]^{-1} = [R_{qq}]^{-1} \frac{1}{v^2}$$

For practical purpose v can be approximated to be 1.0 per unit to make the weight independent of state variable without any substantial effect on the estimation. Equation (A.26) is the result of this approximation.

$$[R'_{qq}] = [R_{qq}] \quad (\text{A.26})$$

3) For line flow in any branch $i-k$ the assumption

$$\cos \theta_{ik} = 1.0 \quad (\text{A.27})$$

$$g_{ik} \sin \theta_{ik} \ll b_{ik} \cos \theta_{ik} \quad (\text{A.28})$$

and for active power flow a further approximation

$$v_i v_k \approx 1.0 \quad (\text{A.29})$$

are made.

4) Series resistance of a line $i-k$ is neglected compared to its series reactance to obtain b_{ik} as

$$b_{ik} = \frac{1}{X_{ik}} \quad (\text{A.30})$$

Table A.2 shows the state vector independent elements of the

Jacobian matrices $[H_{pp}]$ and $[H'_{qq}]$.

In the computations of active measurement residuals $[\Delta Z_p]$, none of the simplifications are applied, while for reactive measurement residuals

$$[\Delta Z'_q] = [Z'_q] - [h'_q(\theta, v)]$$

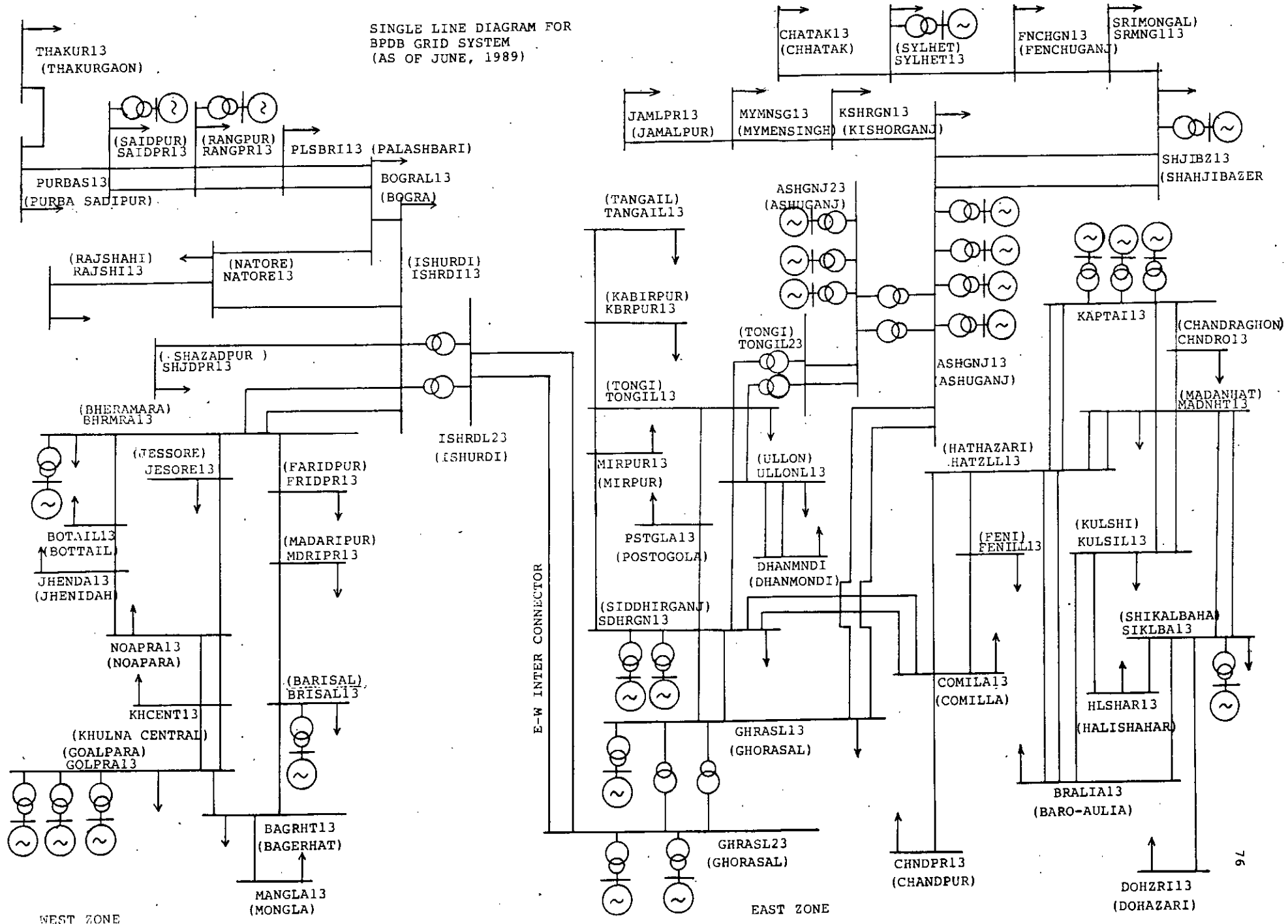
the simplification (1) is applied. Also, the Taylor's expansion, truncated after the third term for sine and cosine functions and the exact expressions for g_{ik} and b_{ik} are used for both the vectors $[\Delta Z_p]$ and $[\Delta Z'_q]$ to improve accuracy and save computation time.

Table A.2: State vector independant elements of active and reactive Jacobian matrices.

Active Jacobian $[H_{pp}]$	$\frac{\partial P_{ik}}{\partial \theta_i} = -b_{ik}$	$\frac{\partial P_{ik}}{\partial \theta_k} = b_{ik}$
	$\frac{\partial P_i}{\partial \theta_i} = \sum_{k \in \alpha_i} \epsilon_{\alpha_i} - b_{ik}$	$\frac{\partial P_i}{\partial \theta_k} = b_{ik}$
Reactive Jacobian $[H'_{qq}]$	$\frac{\partial Q'_{ik}}{\partial v_i} = -b_{ik}^{-s}$	$\frac{\partial Q'_{ik}}{\partial v_k} = b_{ik}$
	$\frac{\partial Q'_i}{\partial v_i} = -\sum_{k \in \alpha_i} (b_{ik} + s)$	$\frac{\partial Q'_k}{\partial v_k} = b_{ik}$
	$\frac{\partial v_i}{\partial v_i} = 1.0$	

APPENDIX B
BPDB GRID SYSTEM DATA

SINGLE LINE DIAGRAM FOR
BPDB GRID SYSTEM
(AS OF JUNE, 1989)



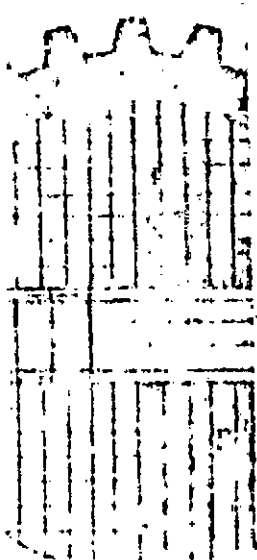
NODES SYSTEM
 BASE MVA = 100
 REFERENCE NODE IS 0P51U210

NODE NO.	NODE NAME	SPECIFIED VOLTAGE P. J.	SPECIFIED GENERATION		SPECIFIED LOAD	
			P	Q	P	Q
			MW	MVAR	MW	MVAR
1	KPTI2X40	1.030	30.0	55.0	1.5	0.75
2	KPTI1X50	1.030	40.0	30.0	0.8	0.40
3	KPTI2X50	1.030	25.0	65.0	1.5	0.75
4	SKLSTX3M	1.030	75.0	55.0	5.2	2.60
5	SDSH3X10	1.030	20.0	15.0	1.4	0.70
6	SDSH3X33	1.030	25.0	80.0	3.0	1.50
7	ASSN2X54	1.030	60.0	40.0	4.2	2.10
8	ASSN1U60	1.030	55.0	40.0	2.0	1.00
9	ASSN1X30	1.030	30.0	20.0	2.0	1.00
10	ASSN2U60	1.030	00.0	00.0	0.0	0.00
11	ASCIU150	1.030	130.0	100.0	11.0	5.40
12	ASS2U150	1.030	130.0	100.0	11.0	5.40
13	ASS3U150	1.030	130.0	100.0	11.0	5.40
14	GRSL2X55	1.030	25.0	75.0	7.5	3.70
15	GRS1U210	1.050	210.0	170.0	15.0	7.00
16	GRS2U210	1.050	150.0	140.0	15.0	7.00
17	S4J3ZGT	1.030	00.0	00.0	0.0	0.00
18	SYLH1X20	1.030	20.0	17.0	0.5	0.25
19	KHL1X110	1.030	40.0	40.0	7.0	3.40
20	KHLN1X50	1.030	00.0	00.0	0.0	0.00
21	KHLN3APP	1.030	20.0	15.0	0.8	0.40
22	GRSL2X20	1.030	00.0	40.0	1.5	0.75
23	GRRA3X20	1.030	00.0	45.0	1.4	0.70
24	SOPR1X20	1.030	12.0	15.0	1.0	0.50
25	RNSP1X20	1.030	20.0	15.0	1.0	0.50
26	KAPTAL13				0.0	0.00
27	CHNDRI13				18.9	8.90
28	MAJNHT13				50.0	24.20
29	SIKL1A13				6.2	3.00
30	HATELL13				0.0	0.00
31	DDHZRI13				14.8	7.20
32	KULSIL13				61.7	29.90
33	ILSHAR13				40.0	19.4
34	BRALIA13				35.0	17.0
35	FEHILL13				31.3	15.2
36	COMILA13				34.9	16.9
37	CHDPR13				9.7	4.7
38	SJIRGN13				72.8	35.3
39	ULLONL13				73.7	35.7
40	DHANNDI				38.2	18.5
41	TONGIL13				77.2	37.4
42	MIRPUR13				61.0	29.6
43	PSTGLA13				70.7	34.3
44	KARPUR13				15.0	7.8
45	INSAIL13				22.4	10.9
46	SHRASL13				44.7	21.7
47	ASHGUL13				21.5	10.4

48	SHJIBZ13	7.0	3.4
49	SRMNDL13	10.0	4.8
50	FNCHG113	7.0	3.4
51	SYLHET13	19.7	9.6
52	CHATAK13	7.2	3.5
53	KSHRGM13	12.3	6.0
54	MYINSG13	24.3	11.8
55	JANLPR13	13.6	6.6
56	GOLPRA13	1.8	.9
57	KACEHT13	70.4	34.1
58	MOAPRA13	10.2	4.9
59	JESORE13	35.0	17.0
60	JHENDAI3	18.0	9.2
61	BJTAL13	21.9	10.6
62	BHRMRA13	16.0	7.8
63	ISHRDI13	26.4	12.8
64	NATORE13	8.9	4.3
65	RAJSHI13	29.7	14.4
66	SHJOPR13	10.8	5.2
67	BOGRAL13	33.0	16.0
68	PLSBRI13	8.7	4.2
69	RAJOPR13	22.8	11.0
70	SAJOPR13	21.0	10.2
71	PURBAS13	13.0	6.3
72	THAKUR13	13.0	6.3
73	BAJRHT13	10.0	4.8
74	MANGLA13	2.0	1.0
75	BRISAL13	22.5	10.9
76	MORIPR13	10.5	5.1
77	FRIDPR13	13.7	6.6
78	ASISHU23		
79	SIRASL23		
80	TONGIL23		
81	LSIRDL23		

LINE NO.	SENDING END	RECEIVING END	SUSCEPTANCE		
			R P.U.	X P.U.	P.U.
1	KAPTAL13	CHNDR113	.0047	.0177	.0041
2	KAPTAL13	MADNHT13	.0226	.0961	.0200
3	KAPTAL13	HATZLL13	.0222	.0258	.0202
4	KAPTAL13	HATZLL13	.0222	.0258	.0202
5	CHNDR113	MADNHT13	.0179	.0683	.0159
6	MADNHT13	SIKLBA13	.0074	.0348	.0085
7	MADNHT13	SIKLBA13	.0074	.0348	.0085
8	MADNHT13	KULSIL13	.0074	.0301	.0061
9	MADNHT13	KULSIL13	.0074	.0301	.0061
01	MADNHT13	HATZLL13	.0050	.0186	.0046
11	MADNHT13	HATZLL13	.0050	.0186	.0046
12	HATZLL13	FEHILL13	.0516	.1712	.0474
13	FEHILL13	COHILA13	.0282	.1066	.0259
14	HATZLL13	COHILA13	.0723	.2958	.0733
15	SIKLABA13	HUGHAD13	.0036	.0223	.0058
16	SIKLABA13	DDIZRI13	.0138	.0717	.0165
17	HATZLL13	BRALIA13	.0070	.0258	.0064
18	HATZLL13	BRALIA13	.0070	.0258	.0064

19	KJLSIL13	HLCHAP13	.0079	.0321	.0065
20	KJLSIL13	BTALIA13	.0075	.0309	.0061
21	CMILAL13	SDIRGN13	.0510	.1888	.0468
22	CMILAL13	SDIRGN13	.0510	.1888	.0468
23	CMILAL13	CINDPR13	.0072	.1431	.0298
24	SDIRGN13	ULLDNL13	.0089	.0363	.0073
25	SDIRGN13	GHRASL13	.0259	.0246	.0236
26	SDIRGN13	GHRASL13	.0258	.0246	.0236
27	ULLDNL13	DHANWIDI	.0031	.0034	.1054
28	ULLDNL13	DHANWIDI	.0031	.0034	.1054
29	JLLDNL13	TONGIL13	.0112	.0427	.0099
30	JLLDNL13	TONGIL13	.0112	.0427	.0099
31	TONGIL13	MIRPUR13	.0089	.0342	.0069
32	TONGIL13	PSTGLA13	.0251	.1026	.0207
33	SDIRGN13	MIRPUR13	.0261	.1234	.0250
34	PSTGLA13	SDIRGN13	.0124	.00550	.0112
35	TONGIL13	KBRPUR13	.0123	.0464	.0109
36	GHRASL13	ASHGHJ13	.0254	.0932	.0232
37	GHRASL13	ASHGHJ13	.0254	.0932	.0232
38	KBRPUR13	TIDAIL13	.0223	.1130	.0265
39	ASHGHJ13	SHJIBZ13	.0302	.1112	.0277
40	ASHGHJ13	SHJIBZ13	.0302	.1112	.0277
41	SHJIBZ13	SRINGL13	.0212	.0806	.0188
42	SRINGL13	FHCOSL13	.0286	.1087	.0254
43	FHCOSL13	SYLHET13	.0133	.0697	.0162
44	SYLHET13	CHATA13	.0187	.0720	.0166
45	ASHGHJ13	KSIRGN13	.0301	.1143	.0269
46	KSIRGN13	HYMNS13	.0343	.1229	.0305
47	HYMNS13	JAHPR13	.0321	.1211	.0284
48	GOLPRA13	KACENT13	.0006	.0018	.0091
49	GOLPRA13	KACENT13	.0006	.0018	.0091
50	KACENT13	NDAPRA13	.0168	.0421	.0109
51	KACENT13	NDAPRA13	.0168	.0421	.0109
52	NDAPRA13	JESURE13	.0124	.0566	.0126
53	NDAPRA13	JHNDIA13	.0543	.1585	.0353
54	JESURE13	BHARMA13	.0361	.0511	.0559
55	JHNDIA13	BTAIL13	.0337	.0293	.0219
56	BTAIL13	BHARMA13	.0175	.0509	.0113
57	BHARMA13	ISIRDI13	.0074	.0212	.0049
58	BHARMA13	ISIRDI13	.0074	.0212	.0049
59	JULPRA13	BAGRHT13	.0336	.1050	.0204
60	BAGRHT13	BRISAL13	.0515	.1609	.0312
61	BAGRHT13	MANGAL13	.0244	.0762	.0148
62	BRISAL13	MORIPR13	.0437	.1294	.0280
63	MORIPR13	FRIDPR13	.0492	.1480	.0320
64	FRIDPR13	BIRNRA13	.0809	.2400	.0521
65	NATURE13	BJGRAL13	.0518	.1511	.0338
66	ISIRDI13	NATURE13	.0275	.0799	.0179
67	ISIRDI13	BJGRAL13	.0723	.2310	.0517
68	NATURE13	RAJSHI13	.0311	.0948	.0193
69	BJGRAL13	PLSBR13	.0382	.1134	.0254
70	BJGRAL13	PLSBR13	.0382	.1134	.0254
71	PLSBR13	RAJSPR13	.0407	.1184	.0265
72	PLSBR13	RAJSPR13	.0407	.1184	.0265
73	RAJSPR13	SATPR13	.0319	.0926	.0208



74	SAIDPP13	SAIDPP13	.0319	.0926
75	SAIDPP13	PURBAS13	.0162	.0474
76	PURBAS13	THAKUR13	.0327	.0983
77	ISIRDL13	SHJDDP13	.0583	.1705
78	ASISNJ23	TONGIL23	.0107	.0507
79	ASISNJ23	TONGIL23	.0107	.0507
80	SHRASL23	ISIRDL23	.0272	.1304
81	SHRASL23	ISIRDL23	.0272	.1304
82	KPTI2X40	KAPTAL13	.0	.1111
83	KPTI1X50	KAPTAL13	.0	.2260
84	KPTI2X50	KAPTAL13	.0	.0880
85	SKLSTX34	SIKLAB13	.0	.0876
86	SJG43X10	SDIRGN13	.0	.1271
87	SJG43X33	SDIRGN13	.0	.1000
88	ASGN2X54	ASISNJ13	.0	.1100
89	ASGN1X50	ASISNJ13	.0	.1700
90	ASGN1X30	ASISNJ13	.0	.3560
91	ASGN2U50	ASISNJ13	.0	.1700
92	ASGN1U50	ASISNJ23	.0	.0700
93	ASGN2U50	ASISNJ23	.0	.0700
94	ASGN3U50	ASISNJ23	.0	.0700
95	GRSL2X55	SHRASL13	.0	.0835
96	GRS1U210	SHRASL23	.0	.0450
97	GRS2U210	SHRASL23	.0	.0450
98	SHJ3ZST	SHJIDZ13	.0	.0800
99	SYL4IX20	SYLHET13	.0	.4428
100	KAL1X110	GOLPRA13	.0	.0950
101	KAL1X50	GOLPRA13	.0	.1486
102	KALNB4PP	GOLPRA13	.0	.2243
103	BRSL2X20	BRISAL13	.0	.1500
104	SHRA1X20	SHRARA13	.0	.1405
105	SUPR1X20	SAIDPP13	.0	.4430
106	RNDP1X20	RANDPP13	.0	.5357
107	ASISNJ23	ASISNJ13	.0	.1057
108	ASISNJ23	ASISNJ13	.0	.1057
109	SHRASL23	SHRASL13	.0	.0774
110	SHRASL23	SHRASL13	.0	.0774
111	TONGIL23	TONGIL13	.0	.0577
112	TONGIL23	TONGIL13	.0	.0577
113	ISIRDL23	ISIRDL13	.0	.0577
114	ISIRDL23	ISIRDL13	.0	.0577

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APPENDIX C
SAMPLE RESULTS

APPENDIX C.1
SAMPLE OUTPUT FOR STATE
ESTIMATION IN PRESENCE
OF RANDOM ERRORS IN THE MEASUREMENTS

FAST DECOUPLED STATIC STATE ESTIMATOR
RESULTS FOR A CASE OF RANDOM NOISE CORRUPTED MEASUREMENTS IN
A 3000 BUS SYSTEM

SIMULATION NUMBER= 7

GENERAL INFORMATION

NO. OF ITERATIONS TO CONVERGE = 10.00
TOLERANCE = 0.00000100
TOTAL ACTIVE MEASUREMENTS = 131
TOTAL REACTIVE MEASUREMENTS = 156
TOTAL NO. OF NODES IN THE SYSTEM = 31
TOTAL NO. OF STATE VARIABLES = 160
REDUNDANCY RATIO = 1.79375

ESTIMATED VOLTAGES AND PHASE ANGLES AT NODES

NODE	NODE NAME	ESTIMATED VOLT (P.U.)	ESTIMATED THETA (DEG)	TRUE VOLT (P.U.)	TRUE THETA (DEG)	(EST.-TRUE) VOLT (P.U.)	(EST.-TRUE) THETA (DEG)
1	KPT12X40	1.02207467	3.1432695	1.02292273	3.0545053	-0.00092506	0.08876419
2	KPT11X50	1.03178501	3.5127364	1.02292273	3.1313076	0.00178528	0.38448877
3	KPT12X50	1.03088284	2.3262370	1.02292273	2.7727127	0.00088310	0.11657429
4	SKLSTX04	1.03125000	3.7639616	1.02292273	0.6173577	0.00125027	0.14550393
5	SDGN3X10	1.03084853	-3.3333468	1.02292273	-3.4835752	0.00084877	0.14533043
6	SDGH3X33	1.02926540	0.3237728	1.02292273	0.4832414	-0.00073433	-0.14446855
7	ASS12X64	1.03000541	4.1440124	1.02292273	4.3271322	0.00000658	0.13625521
8	ASS11J50	1.02922705	5.2243212	1.02292273	5.2020382	-0.00077248	0.02223301
9	ASS41X30	1.03045177	0.0385223	1.02292273	5.4782112	0.00045204	-0.33951187
10	ASS42U50	1.03136763	0.0263843	1.02292273	0.26681557	0.00136099	-0.04176997
11	ASS10I50	1.02162652	1.3755141	1.02292273	0.4382865	-0.00131321	-0.05247234
12	ASS20I50	1.03083023	3.3531554	1.02292273	0.4382865	0.00088120	-0.08193111
13	ASS30I50	1.03036487	2.5020394	1.02292273	0.4382865	0.00034714	0.06394386
14	GRSL2X55	1.02966492	1.7127747	1.02292273	1.5726938	-0.00033655	0.20357320
15	GRS10210	1.05000012	0.0000000	1.05000012	0.0000000	0.00000000	0.00000000
16	GRS20210	1.04911222	-0.6521322	1.05000012	-0.8324414	-0.00088787	0.17430323
17	SHJ35T	1.03011513	-1.3227352	1.02292273	-0.0824265	0.00011532	0.23774137
18	SYL41X20	1.02805763	2.4522221	1.02292273	2.6521425	-0.00193405	0.22994965
19	K4L1X10	1.03013134	-35.9132124	1.02292273	-35.5022155	0.00013161	0.08709717
20	K4L11X51	1.03042723	-37.1272221	1.02292273	-37.2741241	0.00042420	0.13632202
21	K4L18X22	1.02942467	-34.6523375	1.02292273	-34.8372211	-0.00057507	0.17828369
22	BRSL2X20	1.03097057	-33.7535542	1.02292273	-33.0817261	0.00097084	0.32517187
23	B1RA3X20	1.03149128	-25.7434447	1.02292273	-26.1620411	0.00142155	0.44259644
24	SDR1X20	1.02714532	-34.3001253	1.02292273	-34.4029846	-0.00028543	0.02265930
25	RNSP1X20	1.02251422	-32.5132514	1.02292273	-32.2047852	-0.00048851	-0.61856628
26	KAPT113	1.00154202	-1.7222213	1.02292273	-1.7225405	-0.00055625	0.22230871
27	CHNR013	0.22555886	-1.7222171	0.22617322	-2.1562433	-0.00061417	0.22722626
28	MADNHT13	0.23223681	-2.7032220	0.28232540	-2.2179225	-0.00009359	0.21756258
29	SIKLG13	0.23522430	-2.5033210	0.28524354	-2.8345089	-0.00022223	0.26511710
30	HATZLL13	0.23263246	-2.2021745	0.28310635	-2.7012947	-0.00055391	0.19612026
31	DJHZRI13	0.27358751	-3.2630027	0.27935785	-3.3932438	0.00022966	0.12724304
32	KULSILL3	0.27743344	-3.1233227	0.27696376	-3.3205995	0.00046968	0.21591990

33	HLSHAR13	0.07910572	-3.1662950	0.07783235	-3.4140544	0.00025537	0.24775940
34	BRALIA13	0.97860628	-2.3613932	0.07891773	-3.0487324	-0.00031149	0.18685627
35	FENILL13	0.95135444	-5.0114231	0.04994847	-6.0605421	0.00150597	0.04911900
36	CHILAL13	0.05522557	-5.0271262	0.01493324	-5.9954790	0.00105263	0.10834408
37	CHNDPR13	0.94342422	-5.4525872	0.01231727	-6.8345013	0.00067755	-0.02518654
38	SDHRG13	0.97328832	-4.6217312	0.07341317	-4.7816095	0.00057572	0.08987808
39	ULLDNL13	0.05641635	-5.1702312	0.01532272	-5.2772255	0.00047114	0.10529453
40	DANMNDI	0.05461702	-5.2044273	0.05455164	-5.3093653	0.00005539	0.10473824
41	TONGIL13	0.05911962	-4.0922227	0.05335301	-4.1877251	0.00025751	0.10745239
42	MRPURI13	0.04922575	-5.1313322	0.01297711	-5.2355682	0.00019924	0.10119915
43	PSTGLA13	0.05332332	-4.5210133	0.05228652	-4.7055454	0.00040680	0.08373155
44	KBRPRI13	0.04622762	-5.0452732	0.04542272	-5.1922666	-0.00022203	0.10275550
45	TNDAIL13	0.02133222	-5.5127172	0.02064337	-5.5563358	-0.00123732	0.14356225
46	JRASL13	1.00725555	-2.3371271	1.01571232	-2.4645185	0.00145675	0.07741642
47	ASHGNJ13	1.01358202	0.0222422	1.01707642	0.9681557	0.00151253	0.03140950
48	SHJIBZ13	1.01723222	-0.0222122	1.01707622	-0.0204225	0.00078459	-0.00541687
49	JMNGL13	1.00533222	-1.0332222	1.00533222	-1.0782222	0.00070667	0.03471555
50	FCHGN13	0.02092222	-1.9322222	0.02092222	-1.9122222	0.00043565	0.07951355
51	SYLHET13	0.02092222	-2.0722222	0.02092222	-2.1342222	0.00067937	0.11012173
52	CHATAK13	0.02092222	-2.3322222	0.02092222	-2.4572222	0.00026166	0.12303225
53	KSARGN13	0.02722222	-1.9122222	0.02722222	-2.0352222	0.00132207	0.11954980
54	MYANSG13	0.04322222	-4.5722222	0.04322222	-4.7982222	0.00365247	0.21981232
55	JAMLPRI13	0.03022222	-5.0222222	0.02222222	-5.7662222	0.00243092	0.24306202
56	SOLPRA13	0.08222222	-37.1122222	0.02362222	-37.2742222	-0.00055506	0.15052246
57	KHEENT13	0.02122222	-37.1142222	0.02261322	-37.2742222	-0.00065024	0.15972200
58	NDAPRA13	0.06822222	-35.2222222	0.02154222	-35.3832222	-0.00044573	0.15110222
59	JESURE13	0.05222222	-35.2152222	0.05171322	-35.3852222	0.00049435	0.15979280
60	JHENDAL13	0.04222222	-31.0622222	0.04322222	-32.2471313	-0.00087221	0.27822266
61	BOTAIL13	0.04454522	-29.2322222	0.04437422	-29.5531551	0.00017130	0.32398287
62	SHYRA13	0.05633222	-25.7722222	0.05687367	-26.0745627	0.00095844	0.30038452
63	ISHROI13	0.05702222	-24.3322222	0.05567322	-24.6192222	0.00140321	0.28208223
64	NATORE13	0.02122222	-27.6222222	0.02124222	-27.2574222	0.00190222	0.55747437
65	RAJSHI13	0.02502222	-29.6222222	0.02422222	-29.2243774	0.00093637	0.51811219
66	SADPR13	0.04512222	-28.6122222	0.04252222	-25.6605222	0.00264454	0.04310609
67	BJSRAL13	0.02092222	-31.5722222	0.02517722	-32.0422222	0.00181603	0.45040344
68	PLSBR13	0.02092222	-32.7422222	0.02442222	-34.2344222	0.00220418	0.48585510
69	KANDPR13	0.02092222	-34.5722222	0.02092222	-35.1611533	0.00153222	0.42664356
70	SAIDPR13	0.02092222	-36.2222222	0.02342222	-37.4042222	0.00053787	0.48365784
71	PURBAS13	0.02092222	-37.7122222	0.02092222	-38.1522222	0.00096786	0.42858287
72	THAKUR13	0.02092222	-38.5722222	0.02092222	-39.0563222	0.00258722	0.37638255
73	BAGSHI13	0.02722222	-34.1222222	0.02722222	-34.3673722	-0.00053734	0.17515564
74	MANGLA13	0.02722222	-37.2522222	0.02722222	-37.4544373	-0.00122315	0.12079520
75	BRISAL13	0.02722222	-38.5722222	0.02722222	-38.9532222	0.00030322	0.29960532
76	MDRIPR13	0.02722222	-39.4422222	0.02722222	-39.8755122	0.00123751	0.43153381
77	FRIDPR13	0.04622222	-33.0722222	0.04622222	-33.5071554	-0.00015895	0.43254282
78	ASHULL23	1.01261132	0.0222222	1.01172222	0.3852222	0.00101230	0.03776741
79	JRASL23	1.02092222	-3.0222222	1.02092222	-4.0673666	0.00054264	0.12758827
80	TONGIL23	0.02722222	0.0222222	0.02722222	0.3112222	0.00012656	0.05322450
81	ISHROL23	0.02722222	-17.0222222	0.02722222	-18.0742222	-0.00039970	0.38414001

MEASUREMENT ESTIMATES AND COMPUTED RESIDUALS

MEASUREMENTS INCLUDE COMPLEX INJECTION AND FLOW VOLTAGES
 LISTED FOR TYPE AND UNITS OF MEASUREMENTS:
 TYPE 1 VOLT(V) IN DB

TYPE 2: REACTIVE INJECTION (P) AT A NODE IN M.W.
TYPE 3: REACTIVE INJECTION (D) AT A NODE IN H.V.
TYPE 4: REACTIVE LINE FLOW (PL) IN MW
TYPE 5: REACTIVE LINE FLOW (D) IN MW

MEAS. NO.	MEAS. TYPE	NODE NO.	NODE NAME	TRUE VALUE	SPECIFIED VALUE	ESTIMATED VALUE	SP.-EST. VALUE	SP.-TRUE VALUE	EST.-TRUE VALUE
1	2	27	CHNDPR13	-18.322263	-17.25050	-18.43223	0.481724	0.949466	0.4677415
2	2	33	HLSHAP13	-32.222262	-32.24455	-32.75478	0.518231	1.053410	0.4351795
3	2	35	FMHILL13	-31.222273	-32.42895	-32.25078	2.531834	-1.128976	-1.6609114
4	2	40	CHANNINI	-32.122297	-32.34214	-32.25566	-0.335493	-2.142160	-1.7566681
5	2	47	MIRAJ13	-62.222262	-61.45164	-61.26231	-0.182137	-0.451475	-0.2623377
6	2	47	OSIGL413	-72.522265	-72.47744	-72.45025	-0.328223	-1.779507	-1.7502785
7	2	44	KRPPUR13	-15.222220	-16.24317	-16.72263	-0.113534	-0.943185	-0.7296503
8	2	49	CHNGEL13	-2.222273	-2.23497	-2.24439	-0.085765	0.055915	0.1516812
9	2	50	ENH4513	-6.222233	-7.44537	-7.19251	-0.253856	-0.445379	-0.1925230
10	2	53	KSRAS13	-12.222291	-13.34299	-12.64343	-0.705358	-1.049310	-0.3434420
11	2	54	MYNHS013	-24.222293	-22.77725	-22.48523	0.117278	1.522240	1.4047623
12	2	59	JTSOR13	-34.222262	-35.22265	-35.11609	0.397440	-0.028675	-0.1161158
13	2	60	JHEMD13	-18.822263	-19.54309	-19.83219	0.299106	0.356876	0.0677705
14	2	61	CHTAL13	-21.222263	-21.15174	-20.88817	-0.283545	0.749223	1.0317678
15	2	71	PURDAS13	-12.222293	-12.74436	-14.00136	1.257007	0.255632	-1.0013752
16	2	76	MDRIPR13	-10.422299	-7.22237	-8.50620	0.512927	2.506714	1.9937868
17	2	77	FRIDPR13	-13.622290	-13.46393	-13.35125	-0.112283	0.235457	0.3487408
18	4	26	KAPTAT13	41.722465	42.00257	41.34903	0.659549	0.215102	-0.4434466
19	4	26	KAPTAT13	26.622563	27.12364	26.63410	0.489521	0.425064	-0.0644565
20	4	26	KAPTAT13	71.722727	72.50621	72.29614	0.210075	1.148491	0.9384155
21	4	26	KAPTAT13	71.322727	72.38473	72.22614	0.938576	1.027011	0.9384155
22	4	27	CHNDPR13	22.722476	21.22744	22.81407	-1.539627	-1.513021	0.0266075
23	4	28	MADNHT13	-2.322700	-2.74464	-3.55009	0.515423	-1.534961	-2.2503853
24	4	28	MADNHT13	-6.322700	-2.10565	-2.56008	-0.545573	-2.795958	-2.2503853
25	4	28	MADNHT13	25.242705	23.10305	24.84018	-1.723122	-2.111738	-0.3886163
26	4	28	MADNHT13	25.242725	25.41237	24.85018	0.557186	0.170570	-0.3886163
27	4	28	MADNHT13	-12.362736	-12.52524	-16.94287	-0.582170	1.842343	2.4245138
28	4	28	MADNHT13	-12.362736	-12.27327	-16.94237	1.243399	3.493404	2.4245138
29	4	30	HATZLL13	31.222745	21.24242	32.15542	0.704132	1.648384	0.2447455
30	4	30	HATZLL13	-0.662752	-2.24360	-1.40467	-0.532109	-1.376109	-0.7369995
31	4	30	HATZLL13	20.012463	20.54040	20.35123	0.125564	0.522233	0.3333688
32	4	29	SEKPA13	35.112296	33.26244	37.04042	1.227021	2.157132	0.9301186
33	4	29	SEKPA13	14.222122	17.22207	17.82013	-0.090052	2.777383	2.9779434
34	4	30	HATZLL13	25.122745	25.40272	26.50668	0.301313	0.609254	0.3079414
35	4	30	HATZLL13	25.122745	25.12272	25.47668	0.620037	0.927979	0.3079414
36	4	32	KJLSIL13	1.952133	3.21241	2.52447	0.618038	-0.747552	-1.3656616
37	4	32	KJLSIL13	-12.222432	-14.12241	-13.18073	-0.232685	1.155013	2.0956993
38	4	36	CHHIL13	-12.222432	-12.22200	-12.65296	-0.262935	-0.580303	0.1897321
39	4	36	CHHIL13	-12.222432	-11.72212	-12.65296	0.263270	1.257702	0.1897321
40	4	36	CHHIL13	0.722222	11.10520	11.10521	-0.000017	1.325270	1.3259888
41	4	38	CHHIL13	32.142253	21.22242	31.74057	0.237220	-0.251352	-0.3227803
42	4	38	CHHIL13	-47.122212	-47.23271	-47.22422	-1.240222	-1.350003	-0.1097798
43	4	38	CHHIL13	-47.122212	-46.25231	-47.22422	3.228574	2.228824	-0.1097798
44	4	39	CHHIL13	12.122224	22.12223	12.22243	0.121353	1.225046	0.3834832
45	4	39	CHHIL13	12.122224	12.22212	12.22243	-1.174028	-0.220514	0.9934832
46	4	39	CHHIL13	-42.022345	-32.22285	-42.12212	0.354615	0.431293	-0.1233220
47	4	39	CHHIL13	-42.022345	-42.22243	-42.12212	-0.515313	-0.632635	-0.1233220

48	4	41	TONGIL13	51.205312	51.39091	52.24327	-0.357558	-0.024515	0.3429532
49	4	41	TONGIL13	5.791113	9.21225	7.11650	-1.095759	-1.421153	0.3253937
50	4	38	SHRSON13	9.456635	7.26888	9.38320	-0.314313	0.387746	-0.0734329
51	4	43	POTGLA13	-63.235952	-63.15335	-65.35644	0.198084	-1.222509	-1.4205933
52	4	41	TONGIL13	39.950977	33.04106	39.24942	-0.208354	0.190573	0.3989279
53	4	46	SHRSL13	-63.736026	-61.49949	-63.15422	1.664733	2.276789	0.6120563
54	4	46	SHRSL13	-63.756267	-62.77709	-63.15422	0.375151	0.987207	0.6120563
55	4	44	KBRPUB13	22.607309	22.43281	22.26791	0.164902	-0.175498	-0.3394008
56	4	47	ASHGJ13	15.227587	17.17446	16.64485	0.459617	1.211588	0.7519722
57	4	47	ASHGJ13	15.227587	16.17308	16.64485	-0.459563	0.292408	0.7519722
58	4	48	SHJIN13	23.626201	23.74267	23.96903	-0.125355	-0.884133	-0.7577777
59	4	47	SHHGL13	13.482241	13.57715	13.83259	-0.212538	-0.812585	-0.6001472
60	4	50	SYLHET13	7.174047	6.02227	5.43471	0.288218	-0.501139	-0.7803567
61	4	51	SYLHET13	7.211424	7.05246	7.05949	-0.000029	-0.142031	-0.1420021
62	4	47	ASHGJ13	51.873273	42.74456	50.77571	-1.531148	-2.634441	-1.1032934
63	4	52	KHCH13	31.477814	31.70139	37.20274	1.426554	0.000084	-1.4164801
64	4	54	MYN13	13.472597	13.77732	13.72442	-0.147028	-0.102371	0.0447273
65	4	55	GOLPRA13	16.910010	16.22322	17.10205	-0.178754	0.413286	0.5920410
66	4	55	GOLPRA13	16.910010	17.22322	17.10205	0.127202	0.789243	0.5920410
67	4	57	KHCH13	-15.474227	-15.20255	-13.81761	0.523055	0.385379	-0.1426756
68	4	57	KHCH13	-15.474227	-13.12207	-13.81761	-0.311470	-0.454145	-0.1426756
69	4	59	MDARA13	-14.157752	-15.34255	-15.87153	0.521981	-0.126814	-0.7187963
70	4	59	MDARA13	-11.922432	-12.53171	-12.81135	0.172635	-0.722265	-0.7019017
71	4	52	JFSORE13	-51.445203	-50.45572	-52.28976	1.803033	0.982491	-0.8205533
72	4	52	JHEUN13	-51.851262	-54.21742	-52.75125	-1.465150	-2.355331	-0.3701715
73	4	51	BTFAL13	-74.210342	-74.32522	-74.72744	0.590515	0.703133	0.1125230
74	4	52	SHRRA13	-26.256475	-23.27371	-25.50296	2.529238	2.982752	0.3535151
75	4	52	SHRRA13	-26.256475	-24.13225	-25.50296	-2.429297	-2.275782	0.3535151
76	4	56	GOLPRA13	17.412457	16.30333	17.15306	-0.349700	-0.615076	-0.2653956
77	4	73	BAGR413	5.312157	1.53173	3.28554	-0.454761	-1.780367	-1.3256073
78	4	73	BAGR413	2.001139	1.44333	1.94339	-0.000063	-0.057355	-0.0577927
79	4	75	BRICAL13	-13.705217	-12.15542	-20.43809	0.281323	-1.450777	-1.7321701
80	4	76	MOR13	-22.526427	-31.02169	-22.256439	-1.657211	-1.425189	0.2320230
81	4	77	FRID13	-43.254732	-44.04124	-43.39546	-0.245502	-0.373393	0.5721092
82	4	64	NATDR13	33.560465	37.03200	36.64025	0.401753	1.501536	1.0997829
83	4	63	ISH13	76.426777	71.22247	71.47702	0.345450	-4.604303	-4.9497538
84	4	61	ISH13	20.271728	23.23777	22.57700	-0.732228	-1.712023	-0.2727955
95	4	64	NATDR13	30.115204	31.12247	31.12247	0.000000	1.073461	1.0734615
85	4	67	BOS13	25.273522	24.15331	24.72771	-0.544415	-0.920205	-0.2757907
87	4	67	BOS13	25.273522	25.34471	24.72771	0.544189	0.369398	-0.2757907
88	4	63	PLSD13	21.328223	21.75692	20.63130	1.123592	1.355673	0.2330899
89	4	62	PLSD13	21.328223	19.50267	20.63130	-1.123635	-0.920547	0.2330899
90	4	52	SHRRA13	11.275927	11.12445	11.64072	0.553667	0.938475	0.3348076
91	4	59	SHRRA13	11.275927	11.27712	12.64079	-0.553673	-0.158865	0.3348076
92	4	70	SATDR13	25.747574	22.42252	27.52796	1.921542	3.181227	1.2603974
93	4	71	SHRRA13	11.275927	11.27712	13.33376	-1.566308	-1.315587	0.2507210
94	4	63	ISH13	12.871201	12.72721	12.72722	-0.000011	1.846015	1.8460274
95	4	78	ASHGJ13	122.516223	152.76285	129.50859	3.261220	3.252997	-0.0083923
96	4	73	ASHGJ13	122.516223	122.37132	122.50859	-1.137256	-1.145647	-0.0083923
97	4	72	SHRRA13	122.622475	125.35653	125.50005	-0.145530	-3.253936	-3.1084061
98	4	72	SHRRA13	122.622475	126.44562	125.50005	0.145626	-2.952782	-3.1084061
99	4	1	KATI13	20.122200	24.21227	24.23597	0.000000	-2.253233	-2.2532332
100	4	2	KATI13	20.122200	20.56962	20.55962	-0.000000	1.369648	1.3696547
101	4	3	KATI13	21.37571	21.37571	21.37571	-0.000000	-2.124154	-2.1241484
102	4	4	SKLSTX13	52.722281	57.46422	57.46422	0.000000	-2.335599	-2.3355894

103	4	5	SSS13X10	12.522284	12.421827	12.421866	0.000017	0.821983	0.8219659
104	4	6	SSS13X13	21.222293	21.222293	21.222293	0.000023	-4.094977	-4.0950232
105	4	7	SSS13X16	55.722266	57.57710	57.57710	0.000017	1.937223	1.9372059
106	4	8	SSS13X19	52.222247	52.222247	52.222247	0.000023	-0.059384	-0.0594079
107	4	9	SSS13X22	27.222278	26.15325	26.15325	0.000000	-1.846653	-1.8466530
108	4	10	SSS13X25	0.000000	-0.77926	-0.77926	0.000000	-0.789260	-0.7892604
109	4	11	SSS13X28	118.222281	116.27153	116.27153	0.000000	-2.627944	-2.6279449
110	4	12	SSS13X31	118.222281	116.11492	116.11492	0.000000	-2.384573	-2.3845787
111	4	13	SSS13X34	118.222281	112.222266	112.222266	0.000000	0.339184	0.3391844
112	4	14	SSS13X37	27.422287	26.222281	26.222281	-0.000000	2.926911	2.9269167
113	4	15	SSS13X40	162.222281	164.45222	164.45222	0.000000	-5.227083	-5.2270899
114	4	16	SSS13X43	136.222266	136.222266	136.222266	0.000000	1.204201	1.2042015
115	4	17	SSS13X46	0.000000	0.000000	0.000000	0.000000	0.297315	0.2973157
116	4	18	SSS13X49	12.422287	22.222281	22.222281	0.000000	0.738167	0.7381619
117	4	19	SSS13X52	32.222278	31.51666	31.51666	0.000041	-1.393066	-1.3931072
118	4	20	SSS13X55	0.000000	-0.28788	-0.28788	0.000017	-0.287866	-0.2878852
119	4	21	SSS13X58	16.192214	12.222266	12.222266	0.000017	0.116227	0.1162112
120	4	22	SSS13X61	-1.000011	-1.12228	-1.12228	-0.000008	0.310225	0.3102336
121	4	23	SSS13X64	-1.222279	0.34222	0.34222	0.000013	1.740173	1.7401648
122	4	24	SSS13X67	10.222266	2.222266	2.222266	-0.000011	-1.708298	-1.7082863
123	4	25	SSS13X70	13.222279	15.70311	15.70311	-0.000011	-3.296834	-3.2968216
124	4	26	SSS13X73	43.222212	46.41131	46.41131	-2.401543	-2.571392	0.2301514
125	4	27	SSS13X76	48.222212	50.222266	50.222266	0.847577	1.077888	0.2301514
126	4	28	SSS13X79	-37.222287	-36.33310	-36.33310	-0.157500	0.935847	1.0944481
127	4	29	SSS13X82	-37.222287	-36.01425	-36.01425	0.157553	1.252007	1.0944491
128	4	30	SSS13X85	127.522278	129.222266	129.222266	2.484742	1.380634	-1.4841080
129	4	31	SSS13X88	127.522278	126.41542	126.41542	-0.473976	-1.959084	-1.4841080
130	4	32	SSS13X91	172.222215	172.222215	172.222215	2.489852	5.475902	2.9860497
131	4	33	SSS13X94	172.222215	172.222215	172.222215	-2.489748	0.496101	2.9860497
132	3	27	SSS13X97	-2.222279	-2.222279	-2.222279	0.751239	-0.435125	-1.1863642
133	3	28	SSS13X100	-12.222266	-12.222266	-12.222266	0.267118	0.191835	0.1247168
134	3	29	SSS13X103	-13.222266	-13.222266	-13.222266	0.191861	2.099990	1.9081287
135	3	30	SSS13X106	-13.222266	-13.222266	-13.222266	-0.193905	0.194340	0.3882468
136	3	31	SSS13X109	-22.222278	-22.222278	-22.222278	0.142972	-0.233163	-0.4261434
137	3	32	SSS13X112	-34.222278	-34.222278	-34.222278	-0.237548	0.153452	0.3910005
138	3	33	SSS13X115	-7.222278	-7.222278	-7.222278	0.623309	0.681060	0.0577592
139	3	34	SSS13X118	-6.222278	-6.222278	-6.222278	-0.437563	-0.434121	0.0053819
140	3	35	SSS13X121	-3.222278	-3.222278	-3.222278	0.870994	0.397722	-0.5632717
141	3	36	SSS13X124	-5.222278	-5.222278	-5.222278	-0.494462	-2.354056	-1.8595943
142	3	37	SSS13X127	-11.222278	-11.222278	-11.222278	-0.267755	2.204879	2.0956335
143	3	38	SSS13X130	-12.222278	-12.222278	-12.222278	0.385339	1.258775	1.5724354
144	3	39	SSS13X133	-12.222278	-12.222278	-12.222278	0.378292	-0.759416	-1.1297088
145	3	40	SSS13X136	-12.222278	-12.222278	-12.222278	-0.200034	-1.025547	-0.7975101
146	3	41	SSS13X139	-6.222278	-6.222278	-6.222278	-0.058272	-0.358247	-0.2992749
147	3	42	SSS13X142	-3.222278	-3.222278	-3.222278	0.619236	1.592153	0.9732172
148	3	43	SSS13X145	-3.222278	-3.222278	-3.222278	0.453376	-1.101256	-1.5546436
149	5	25	SSS13X148	22.222278	24.222278	24.222278	1.267154	1.629283	0.4211187
150	5	26	SSS13X151	11.222278	16.222278	16.222278	1.948356	1.422702	-0.5256534
151	5	27	SSS13X154	11.222278	11.222278	11.222278	0.722206	-0.106663	-0.9288741
152	5	28	SSS13X157	11.222278	11.222278	11.222278	-2.044617	-2.873492	-0.8288741
153	5	29	SSS13X160	12.422287	10.222278	10.222278	-1.757984	-2.524685	-0.7668018
154	5	30	SSS13X163	-2.222278	-2.222278	-2.222278	0.536882	0.234785	0.2272927
155	5	31	SSS13X166	-2.222278	-2.222278	-2.222278	-0.111261	0.285941	0.3979027
156	5	32	SSS13X169	11.222278	2.222278	2.222278	0.356686	-1.434827	-1.7615137
157	5	33	SSS13X172	11.222278	2.222278	2.222278	-1.203215	-2.959729	-1.7615137

158	5	25	HADHT13	2.720495	2.37594	2.53755	0.359291	2.106440	1.7471485
159	5	28	HADHT13	0.720405	2.59324	2.53755	0.055690	1.802839	1.7471485
160	5	30	HATZLL13	7.271204	6.35187	6.02518	0.325688	-0.919127	-1.2448149
161	5	34	FONILL13	-5.604174	-6.22236	-5.01536	-0.705995	0.389812	0.5902087
162	5	39	HATZLL13	1.004077	0.50273	0.40530	0.097429	-0.501298	-0.5987279
163	5	39	SKL3117	22.731713	22.79144	21.65583	1.245605	0.170439	-1.0751657
164	5	27	SKL3113	5.823204	4.71152	4.71159	0.000028	-1.091578	-1.0916071
165	5	39	HATZLL13	0.131335	1.37432	4.17760	0.195713	-0.817573	-1.0142860
165	5	39	HATZLL13	0.171337	0.14074	3.17760	-0.035855	-1.051151	-1.0142860
167	5	32	KULSIL13	-3.044127	-2.29779	-2.27193	-0.215867	0.736590	0.9525582
168	5	32	KULSIL13	-0.225229	-0.23480	-0.77405	-0.160740	1.790418	1.2511595
169	5	36	COMILA13	-7.872923	-9.23327	-7.70503	-0.578242	-0.401252	0.1769900
170	5	36	COMILA13	-7.872923	-7.16343	-7.70503	0.541602	0.818572	0.1769900
171	5	36	COMILA13	0.136775	1.34702	1.44794	0.000076	-0.338755	-0.3388315
172	5	38	SO470N13	49.362742	41.25767	41.79310	-0.735414	0.695157	1.4305706
173	5	32	SO470N13	-29.716284	-20.31913	-21.63221	1.313077	0.397223	-0.2158552
174	5	33	SO470N13	-20.716354	-22.71567	-21.63221	-1.084452	-2.000308	-0.9158552
175	5	39	ULL70N13	-0.310107	-1.26700	-0.55252	-0.714481	-0.906899	-0.1924176
176	5	39	ULL70N13	-0.310107	-0.37496	-0.55252	0.178461	-0.013956	-0.1924176
177	5	39	ULL70N13	2.546592	4.16276	2.19552	1.765744	1.615574	-0.3511693
178	5	39	ULL70N13	2.546592	-0.51529	2.19552	-2.711819	-3.062989	-0.3511693
179	5	41	FONILL13	12.741097	13.02477	12.85112	0.153656	0.283688	0.1200319
180	5	41	FONILL13	-6.439114	-9.14746	-6.55530	-1.493157	-1.710354	-0.2171874
181	5	38	SO470N13	15.460119	15.29452	15.78227	0.505258	0.874429	0.3291607
182	5	43	POSTGLA13	-33.217264	-33.27117	-38.74388	-0.227314	-0.058233	0.1690745
183	5	41	FONILL13	17.094421	17.14704	17.28216	1.157979	2.052634	0.8947551
184	5	46	SFRASL13	5.223356	5.51717	5.72871	-0.091512	-0.375166	-0.2846539
185	5	46	SFRASL13	5.223356	6.20285	5.70971	1.203144	0.215491	-0.2846539
185	5	46	KORRUP13	2.354035	2.35460	10.28722	-7.902622	0.018566	0.9211898
187	5	47	ACHONJ13	-5.602551	-5.20122	-5.18203	-0.102186	0.311328	0.4135147
187	5	47	ACHONJ13	-5.602551	-4.27227	-5.18203	0.315754	0.770279	0.4135147
189	5	43	SUJ13713	6.060067	5.50852	7.33304	-0.734454	-0.354470	0.3599839
190	5	49	SMNCL13	3.584851	3.64359	3.75226	1.573544	1.276258	0.4034135
191	5	50	FONCH13	7.634605	1.22201	2.32140	-1.038420	-1.181563	-0.1430776
192	5	51	SYL13113	1.217102	2.52607	2.57594	0.290013	0.613749	0.6137352
193	5	47	ACHONJ13	22.913223	21.98322	23.19410	-1.113074	-0.729799	0.3832757
194	5	43	ACHONJ13	17.934927	15.22336	14.50975	0.783610	-0.570911	-1.3545208
195	5	44	BYNCL13	4.422562	5.46676	5.32966	0.775074	1.046101	0.2700073
195	5	46	SOLO3A17	47.212759	42.19166	49.78349	0.403267	-0.027066	-0.4303336
197	5	46	SOLO3A17	47.212759	42.19271	49.78340	-0.3398571	-0.829005	-0.4303336
199	5	47	KLEF1313	33.013917	33.40333	32.60568	0.345122	0.466882	-0.3782332
199	5	47	KLEF1313	33.013917	31.58052	33.63563	-1.085072	-1.463305	-0.3782332
200	5	43	SO470N13	27.077549	17.73222	37.70211	1.030445	-0.340855	-1.3715200
201	5	38	SO470N13	27.077571	25.17073	27.27725	-1.101517	-0.452034	0.6486833
202	5	39	JUT13713	18.317217	17.59374	18.55728	1.025445	1.270525	0.2440810
203	5	40	JUT13713	18.317217	16.45209	18.23286	-2.517782	-3.164077	-0.6482542
204	5	41	ACHONJ13	3.037245	2.77241	4.05715	-0.277741	-1.829450	-1.5307073
205	5	42	ACHONJ13	27.052154	34.11285	33.69409	0.413752	-1.745225	-2.1540472
205	5	42	ACHONJ13	27.052154	32.23152	33.69409	-0.762414	-2.925462	-2.1540472
207	5	45	SOLO3A17	-1.122371	-0.37721	-1.21462	-0.362586	-0.876564	-0.0310772
208	5	42	SOLO3A17	-3.035009	-5.15073	-4.03249	-1.118244	-1.214823	-0.0965800
209	5	43	SOLO3A17	-3.035009	-0.31197	-2.42553	0.000039	0.898095	0.8980453
210	5	75	ACHONJ17	11.174502	21.00173	21.17033	-0.088560	-0.142722	-0.8541587
211	5	76	ACHONJ17	17.557224	16.84122	18.28301	-1.541929	-0.705421	0.8356094
212	5	77	ACHONJ17	17.557224	7.12254	11.11565	-3.223102	-3.264012	-0.7402036

213	5	34	HAT00113	-3.755237	-4.17729	-3.99423	-0.993747	-1.121747	-0.2379994
214	5	33	IS470113	13.421225	10.53305	12.32122	-0.788182	0.051051	0.8392334
215	5	63	ISH00113	4.221225	4.34322	4.32241	0.017345	0.057567	0.0491840
216	5	34	HAT00113	14.271217	14.11321	14.33323	-0.000017	0.754600	0.7546186
217	5	67	B0500113	-8.072211	-7.25252	-3.75174	0.792151	0.563490	-0.2286613
218	5	67	B0700113	-8.522211	-10.11924	-4.75174	-1.364105	-1.592765	-0.2286613
219	5	68	PL500113	-2.422211	-10.52224	-2.05265	-1.528614	-1.108175	0.4204392
220	5	63	PL500113	-3.422211	-4.12222	-9.05966	0.962275	1.382714	0.4204392
221	5	62	RAN00113	-3.622211	-2.11144	-2.75524	0.640727	1.425401	0.8546020
222	5	69	SA000113	-2.422211	-2.42224	-2.75524	-0.991204	-0.026402	0.8546020
223	5	70	SA000113	12.722211	3.44221	3.53373	-0.091719	-1.307042	-1.2173233
224	5	71	PJR00113	4.222211	3.34215	3.28830	0.073856	-1.464674	-1.5385313
225	5	62	IS470113	21.222112	1.07213	1.37411	0.000021	-1.249673	-1.2497005
226	5	73	AS000113	36.753211	33.00623	34.55233	-0.545050	1.252443	1.7784982
227	5	74	ASH00113	35.753211	31.54437	34.55233	-0.007953	1.790535	1.7984982
228	5	72	GH000113	27.036224	27.72110	25.63055	-3.042454	0.545050	0.5945265
229	5	70	G0000113	25.036224	24.54524	25.63055	1.014685	1.602211	0.5945265
230	5	1	K0112X40	22.122211	22.17847	29.64004	0.519444	-0.009970	-0.5483747
231	5	2	K0111X50	14.422211	14.72526	15.61055	-0.714538	0.475010	1.2025285
232	5	3	K0112X50	36.372211	36.74423	37.23310	-1.120173	0.355537	1.5467167
233	5	4	S0112X34	43.221077	44.45758	55.02113	1.135433	2.555476	1.4200621
234	5	5	S0113X10	47.045123	46.72528	48.23411	-1.509122	-1.212206	0.2882156
235	5	6	S0113X33	61.512211	61.23365	60.75572	3.477248	1.720047	-1.7572012
236	5	7	AS000113	13.722211	9.22222	12.42423	-4.044812	-5.347316	-1.3025045
237	5	8	AS000113	10.111211	7.92222	4.71731	1.270287	-0.125537	-1.3259273
238	5	9	AS000113	5.022211	6.68212	4.52506	1.022070	0.614108	-0.4779629
239	5	10	AS000113	7.822211	7.17222	7.74225	-0.575174	-0.656986	-0.0808120
240	5	11	AS000113	31.222211	32.75222	22.12221	2.577584	-1.082389	-3.6522751
241	5	12	AS000113	31.022211	31.61222	31.42224	0.185740	-0.222015	-0.4077554
242	5	13	AS000113	31.222211	32.52222	30.22224	-0.311825	-1.241765	-0.9099424
243	5	14	AS000113	32.222211	31.12222	30.22221	0.274624	-1.758725	-2.0334892
244	5	15	AS000113	53.622211	53.27222	62.05606	3.220170	1.585429	-1.6347408
245	5	16	AS000113	61.622211	56.42222	58.20414	-1.717427	-4.922938	-3.2735109
246	5	17	AS000113	16.622211	15.42222	15.68677	-0.211762	-1.149215	-0.9381473
247	5	18	S0112X30	2.422211	2.27222	4.39072	0.387829	-0.170355	-0.5581856
248	5	17	K0111X10	22.722211	22.22222	51.61259	-0.719219	0.100070	0.8182893
249	5	20	K0111X30	22.142211	21.22222	32.21026	-0.922707	-0.155735	0.7549720
250	5	21	K0111X50	21.722211	21.22222	21.73512	-0.482954	-0.452397	0.0305653
251	5	22	R0000113	34.722211	35.51222	36.28259	0.234687	0.725537	0.4908502
252	5	23	R0000113	22.722211	22.75222	34.01163	0.253220	1.423075	0.4691958
253	5	24	R0000113	22.722211	22.16622	22.72276	1.323854	0.443762	-0.9500921
254	5	25	R0000113	22.722211	22.25222	23.22227	1.252265	0.522548	-0.7304192
255	5	26	AS000113	-4.022211	-1.74222	-4.42421	2.438837	2.162958	-0.4759486
256	5	27	AS000113	-1.022211	-4.98367	-4.43421	-0.403837	-0.879786	-0.4759486
257	5	28	AS000113	2.322211	24.32444	25.12716	-0.722719	-1.254825	-1.2471730
258	5	29	AS000113	22.222211	22.21122	25.10716	1.104205	-0.137972	-1.2421740
259	5	30	AS000113	22.222211	24.14722	38.05034	-3.713074	-4.042255	-0.3368616
260	5	30	AS000113	22.122211	21.12402	38.05034	3.133737	2.795876	-0.3368616
261	5	31	AS000113	5.722211	2.62222	3.02270	0.587230	-2.036130	-2.6234102
262	5	31	AS000113	5.722211	2.62222	3.02270	-0.234471	-2.857881	-2.6234102
263	1	1	K0112X40	1.022211	1.02222	1.02227	-0.000120	-0.001045	-0.0009251
264	1	2	K0111X50	1.022211	1.02222	1.02227	0.000124	0.001210	0.0017853
265	1	3	K0112X50	1.022211	1.02222	1.02227	0.000287	0.001170	0.0008831
266	1	4	S0112X34	1.022211	1.02222	1.02227	-0.000307	0.000743	0.0012507
267	1	5	S0113X10	1.022211	1.02222	1.02227	0.000433	0.001281	0.0008489

258	1	5	ASSH3X33	1.030000	1.02252	1.02225	-0.000669	-0.001403	-0.0007343
267	1	7	ASSH2X54	1.030000	1.03121	1.03000	0.001203	0.001210	0.0000067
270	1	8	ASSH1150	1.030000	1.02297	1.02222	-0.000253	-0.001025	-0.0007725
271	1	9	ASSH1X30	1.030000	1.03037	1.03045	-0.000143	0.000309	0.0004520
272	1	10	ASSH2J50	1.030000	1.03157	1.03136	0.000213	0.001574	0.0013609
273	1	11	ASSH0150	1.030000	1.02804	1.02868	-0.000641	-0.001955	-0.0013132
274	1	12	ASSH0150	1.030000	1.03043	1.03048	-0.000045	0.000835	0.0008812
275	1	13	ASSH0150	1.030000	1.03043	1.03034	0.000083	0.000430	0.0003471
276	1	14	ASSL2X50	1.030000	1.02252	1.02266	-0.000073	-0.000410	-0.0003366
277	1	15	ASSL1J10	1.030000	1.02272	1.05000	-0.001274	-0.001274	0.0000000
278	1	16	ASSL20210	1.030000	1.02267	1.04211	0.000558	-0.000327	-0.0008879
279	1	17	ASSL2021	1.030000	1.02266	1.03011	0.000145	0.000261	0.0001154
280	1	18	ASSL1X20	1.030000	1.02202	1.02205	-0.000044	-0.001978	-0.0019341
281	1	19	KHL1X110	1.030000	1.03027	1.03013	0.000039	0.000371	0.0001316
282	1	20	KHL11X50	1.030000	1.03020	1.03042	0.000275	0.000204	0.0004282
283	1	21	KHL10120	1.030000	1.02252	1.02242	0.000100	-0.000474	-0.0005751
284	1	22	ASSL2X20	1.030000	1.03020	1.03097	-0.000066	0.000904	0.0009708
285	1	23	BTRAB3X20	1.030000	1.03126	1.03142	-0.000228	0.001262	0.0014915
286	1	24	SSP1X20	1.030000	1.02200	1.02214	-0.000141	-0.002225	-0.0028543
287	1	25	RNSP1X20	1.030000	1.02241	1.02251	-0.000104	-0.000588	-0.0004835

VALUES OF PERFORMANCE INDICES

PERFORMANCE INDEX AS A FUNCTION OF:	COMPUTED VALUE P.U.	THRESHOLD VALUE P.U.
1. (SP.-EST.)	0.34725273	0.50210147
2. (EST.-TRUE)	0.52855310	0.74447260

APPENDIX C.2
SAMPLE OUTPUT FOR STATE
ESTIMATION IN PRESENCE OF
BOTH RANDOM AND GROSS
ERRORS IN THE MEASUREMENTS

FAST DECOUPLED STATIC STATE ESTIMATOR
RESULTS FOR A CASE OF BOTH RANDOM NOISE AND GROSS
ERROR CONTAMINATED MEASUREMENTS IN OPDQ GRID SYSTEM

GENERAL INFORMATION

NO. OF ITERATIONS TO CONVERGE = 20.00
TOLERANCE = 1.00000100
TOTAL ACTIVE MEASUREMENTS = 228
TOTAL REACTIVE MEASUREMENTS = 253
TOTAL NO. OF NODES IN THE SYSTEM = 31
TOTAL NO. OF STATE VARIABLES = 160
REUNDANCY RATIO = 3.00625

		ESTIMATED VOLTAGES AND PHASE ANGLES AT				NODES	
NODE	NODE NAME	ESTIMATED VOLT(P.U.)	ESTIMATED THETA(DEG)	TRUE VOLT(P.U.)	TRUE THETA(DEG)	(EST.-TRUE) VOLT(P.U.)	(EST.-TRUE) THETA(DEG)
1	KPFI2X40	1.02921200	3.1130968	1.02929973	3.0545053	-0.00078773	0.05949116
2	KPFI1X50	1.03039265	3.2432250	1.02929973	3.1313076	0.00039291	-0.08828259
3	KPFI2X50	1.02979279	2.7517991	1.02929973	2.7797127	-0.00020695	-0.02792454
4	SKLSTX81	1.02935123	-0.5214027	1.02929973	-0.6173577	-0.00064850	-0.09595495
5	SDGN3X10	1.03091717	-3.5444431	1.02929973	-3.4836769	0.00091743	-0.06076622
6	SDGN3X33	1.03062248	0.5125297	1.02929973	0.4832414	0.00062275	0.02928829
7	ASSN2X54	1.02993543	4.2293732	1.02929973	4.3271322	-0.00105430	-0.10675907
8	ASSN1U60	1.03084278	5.7985229	1.02929973	5.9023882	0.00084305	-0.10356617
9	ASSN1X30	1.02991953	6.4159985	1.02929973	6.4232112	-0.00018120	-0.01221275
10	ASSN2U60	1.02917471	0.9114993	1.02929973	0.9681557	-0.00082302	-0.08654743
11	ASS1U150	1.03097194	3.5930391	1.02929973	3.4389865	0.00057220	0.06794357
12	ASS2U150	1.03012743	3.4131475	1.02929973	3.4382965	0.00012970	-0.01889896
13	ASS3U150	1.02995632	2.4518333	1.02929973	2.4389865	-0.00004292	0.01346874
14	GRSL2X55	1.02923827	1.5233493	1.02929973	1.5795938	-0.00031144	0.04655552
15	GRS1U210	1.03000012	0.0000000	1.03000012	0.0000000	0.00000000	0.00000000
16	GRS2U210	1.04000000	-0.7799999	1.03000012	-0.8324414	-0.00014496	0.05250645
17	S4J3Z51	1.03000014	-0.1835077	1.02929973	-0.0804765	0.00030041	-0.04301119
18	SYL41X20	1.03106117	2.2736276	1.02929973	2.6591425	0.00106144	0.31348515
19	K4L1X110	1.03015514	-37.2432237	1.02929973	-35.6002155	0.00015640	0.25659180
20	K4L1X60	1.02968393	-35.2423674	1.02929973	-37.2741241	-0.00041580	0.33175659
21	K4L1X300	1.03102741	-34.5137213	1.02929973	-34.9372211	0.00010768	0.32102966
22	33L2X20	1.02777717	-32.4081141	1.02929973	-32.9817261	-0.00222206	0.41757202
23	B4R4X20	1.03073507	-25.2666396	1.02929973	-26.1890411	0.000373528	0.22438049
24	SDR1X20	1.02982453	-34.2362276	1.02929973	-34.4029846	0.00002480	0.10890198
25	K15P1X20	1.03112221	-29.2636757	1.02929973	-29.3047352	0.00112247	0.25910250
26	KAPTAI13	1.02920010	-1.7131122	1.02929973	-1.7225496	0.00010109	-0.02057934
27	C4NDRU13	0.99627511	-2.1771836	0.99617302	-2.1562433	0.00009309	-0.00094032
28	4AJNHT13	0.98249696	-2.2663961	0.98238563	-2.2917926	0.00010866	-0.04651356
29	S1KLBAL13	0.99618731	-2.15712947	0.99606554	-2.2345087	0.00044078	-0.04569530
30	HATZLL13	0.99322463	-2.7421274	0.99319433	-2.7012949	0.00003827	-0.04783249
31	UDZKR113	0.97309373	-3.4131122	0.97238725	-3.3237439	0.00064648	-0.01989460
32	KULSIL13	0.97734079	-3.2667355	0.97696376	-3.3205996	0.00037674	-0.04615593
33	HLS4AR13	0.97783322	-3.4734059	0.97793935	-3.4140644	0.00009387	-0.06436052
34	BXAL1A13	0.97332467	-3.0005362	0.97391772	-3.2487324	0.00009695	-0.04379749

35	FEHILL13	0.26032736	-4.0676727	0.29234847	-6.0605421	-0.00052011	-0.00713062
36	COMILAI3	0.25347494	-6.1304083	0.25423394	-6.0254790	-0.00145900	-0.04292774
37	CINOPR13	0.24122225	-6.2137454	0.24781727	-6.8345013	-0.00159502	0.01945591
38	SDRGR13	0.27375600	-4.7761304	0.27341317	-4.7816095	-0.00004709	0.02752876
39	ULLONL13	0.26520267	-5.2475315	0.25532972	-5.2772255	0.00057955	0.02954388
40	DHANND1	0.25516500	-5.2731440	0.25455154	-5.3093653	0.00061417	0.03120041
41	TNSIL13	0.25247247	-4.1627377	0.27725301	-4.1277251	0.00051941	0.04538727
42	MIRPUR13	0.25026933	-5.1214340	0.24930711	-5.2355680	0.00116122	0.03913403
43	PSTGLA13	0.26309047	-4.8203376	0.26228650	-4.7055454	0.00002388	0.01523781
44	KBRPUR13	0.24753722	-5.1647550	0.24542772	-5.1702656	0.00103748	0.02401151
45	TVJAIL13	0.22747215	-6.8077244	0.22609937	-6.6563358	0.00141078	0.14861107
46	SARASL13	1.00622297	-2.4270940	1.00571980	-2.4645155	0.00051117	0.03751373
47	ASHGNJ13	1.01714397	0.2036633	1.01707649	0.2681557	0.00007248	-0.05949497
48	SHJH213	1.01880162	-0.1112846	1.01709869	-0.0804965	0.00171471	-0.03148800
49	SRNHGL13	1.00706126	-1.0224015	1.00582402	-1.02980206	0.00123787	0.00559902
50	FNCHGN13	0.22321012	-1.2527436	0.22556314	-1.2190426	-0.00028224	-0.03372097
51	SYLHET13	0.22347675	-2.1749315	0.2222415	-2.1940239	0.00055259	0.00902233
52	CHATAK13	0.22952281	-2.6472560	0.22952726	-2.4570141	-0.00097445	0.01975918
53	KSARG13	0.27435217	-1.2511762	0.27586250	-2.0357370	-0.00150335	0.08556090
54	MYNASS13	0.23753201	-4.7201117	0.24032207	-4.7287566	-0.00273707	0.02832794
55	JAILPR13	0.22661536	-5.8223650	0.22248146	-5.7663794	-0.00186610	-0.06098652
56	SOLPRAI3	0.22522497	-37.0724102	0.22362291	-37.2741241	0.00160116	0.20170593
57	KHENT13	0.22421478	-37.0750732	0.22261338	-37.2740734	0.00160140	0.19900513
58	NDAPRA13	0.22031233	-35.1615225	0.22115443	-35.3233618	0.00115490	0.22183229
59	JESSRE13	0.22300702	-37.1575754	0.22171320	-35.3856049	0.00122181	0.21804810
60	JHENDAI3	0.24465934	-32.2126617	0.24323215	-32.2471313	0.00141919	0.03446750
61	BDTAL13	0.24752055	-22.3705070	0.24437462	-22.5631561	0.00321603	0.12245911
62	BHRNRA13	1.25755614	-22.2214167	0.25597367	-22.2745697	0.00159247	0.16615295
63	ISHRU13	0.25733750	-24.4579163	0.25567874	-24.4619076	0.00165915	0.16116333
64	NATDR13	0.22152217	-22.1772941	0.21240397	-22.2574005	0.00212520	0.13009644
65	KAJSH13	1.31220507	-22.5301112	0.22423935	-22.2243774	0.00396794	0.22224561
66	SHJDR13	0.24622746	-25.5532240	0.24250751	-25.6605988	0.00448995	0.10670471
67	BDJRAL13	0.22740273	-31.2317322	0.22051720	-32.0403290	0.00223184	0.10859680
68	PLSBR13	0.22672276	-34.1052413	0.22443275	-34.2344971	0.00222001	0.12855530
69	KALDR13	0.220773178	-35.2612294	0.22677780	-35.1611533	0.00100393	0.02486339
70	SADPR13	0.22425003	-37.2214816	0.22340158	-37.4042072	0.00054847	0.11442566
71	KJASAS13	0.22521220	-32.0722654	0.22220526	-32.1522884	0.00200623	0.10713176
72	FAKU13	1.82520427	-32.8716125	0.22177330	-32.2563293	0.00312527	0.08471680
73	SAGHT13	0.22737636	-32.3216279	0.22805292	-32.3673706	0.00070047	0.27374268
74	MADSLAI3	0.22823111	-33.2212433	0.22737036	-33.4544373	0.00020295	0.24519348
75	BRIGALI3	0.22962067	-32.7424144	0.22913412	-32.2532185	0.00122151	0.21150208
76	DRIPR13	0.25143717	-32.5427112	0.22813251	-32.3753182	-0.00227745	0.23330688
77	FRIDR13	0.22122772	-33.3212283	0.22462344	-33.5071564	0.00093712	0.13078308
78	ASFULL23	1.01256557	1.2271514	1.01157932	1.2526042	0.00026798	-0.00044346
79	CHACGL23	1.02714951	-1.2271745	1.02528567	-4.0673666	-0.00014114	0.03342209
80	TDHIL23	0.22712334	0.1104454	0.22553922	0.1112088	0.00056660	0.00353700
81	ISHRDL23	0.22126613	-17.077736	0.22290452	-17.0743356	0.00088151	0.12676208

MEASUREMENT ESTIMATES AND COMPUTED RESULTS

MEASUREMENTS INCLUDE COMPLEX INJECTION AND FLUX VOLTAGES

LISTED FOR TYPE AND UNIT OF MEASUREMENTS:

TYPE 1 VOLT(V) IN PU

TYPE 4: ACTIVE LINE FLUX (PL) IN MW

TYPE 5: REACTIVE LINE FLUX (QL) IN MVAR

MEAS. NO.	MEAS. TYPE	NODE NO.	NODE NAME	TRUE VALUE	SPECIFIED VALUE	ESTIMATED VALUE	SP.-EST. VALUE	SP.-TRUE VALUE	EST.-TRUE VALUE
1	4	27	CHUDPOL3	-41.637657	0.00000	-39.91622	39.916229	41.687469	1.7712526
2	4	28	MADNHT13	-26.479507	-26.53108	-26.95398	0.432908	-0.051569	-0.4844785
3	4	30	HATZLL13	-70.195007	-72.57386	-71.34857	-1.225297	-2.378862	-1.1535645
4	4	30	HATZLL13	-70.195007	-71.37577	-71.34857	-0.027215	-1.180779	-1.1535645
5	4	28	MADNHT13	-22.656391	-24.24453	-23.71463	-0.529909	-1.587640	-1.0577316
6	4	29	STKLBA13	5.320285	4.77425	5.59751	-1.821360	-1.544032	0.2773285
7	4	27	STKLBA13	6.320285	5.76372	5.59751	0.171381	0.448709	0.2773285
8	4	32	KJLSIL13	-25.189377	-23.23319	-24.98092	1.747727	1.956183	0.2084613
9	4	32	KJLSIL13	-25.189377	-27.75295	-24.98092	-2.771949	-2.563483	0.2084613
10	4	30	HATZLL13	19.336932	20.37635	19.19487	1.191490	0.989419	-0.2020717
11	4	30	HATZLL13	19.336932	20.39467	19.18487	1.209598	1.007525	-0.2020717
12	4	35	FMHLL13	-39.632187	-29.93767	-30.36851	0.528842	0.792521	0.2636790
13	4	36	CMHLL13	9.574957	-0.27072	-0.04577	-0.224944	-0.944777	-0.7198334
14	4	36	CMHLL13	-17.670563	-19.93687	-19.73771	-0.199157	-0.266301	-0.0671446
15	4	37	HLSHA13	-35.042121	-35.91531	-37.27388	1.358556	0.126874	-1.2316818
16	4	31	DHZZT13	-14.799279	-14.57554	-14.19037	-0.345171	0.273430	0.6196022
17	4	34	BRALTA13	-25.146103	-23.47977	-24.93295	1.253175	1.466321	0.2131462
18	4	34	BRALTA13	-25.146103	-26.80064	-24.93295	-1.867637	-1.554541	0.2131462
19	4	37	HLSHA13	-3.057653	-5.18728	-5.08651	-1.600766	-2.729630	-1.1288643
20	4	34	BRALTA13	15.225419	13.73518	15.15264	-1.914556	-2.059221	-0.1427650
21	4	38	SDHRS13	12.952702	13.80265	13.63608	0.117570	0.950945	0.7333755
22	4	38	SDHRS13	12.952702	13.79798	13.63608	0.021828	0.755274	0.7333755
23	4	37	CHIDPR13	-9.639217	-8.71141	-9.15174	0.449329	0.988501	0.5481720
24	4	39	ULLOM13	-31.827624	-30.25434	-31.43910	0.484746	0.933247	0.4485011
25	4	46	BRASL13	47.993178	46.56582	48.24073	-1.675045	-1.332288	0.3427565
26	4	46	BRASL13	47.993178	71.04341	48.24073	7.802457	3.145223	0.3427565
27	4	49	DIAMND1	-11.088745	-12.85517	-11.20220	-0.657274	0.233560	0.8865356
28	4	40	DIAMND1	-11.088745	-12.50975	-11.20220	-1.398556	-0.512021	0.8865356
29	4	41	TUNGIL13	49.201447	49.72435	49.83088	-0.106042	0.523394	0.6294370
30	4	41	TUNGIL13	49.201447	43.00454	49.83088	2.175652	2.905090	0.6294370
31	4	42	MIRPJD13	-51.647372	-51.36325	-51.65053	0.297480	0.280314	-0.0171661
32	4	43	PSTGLA13	-9.770937	-7.97253	-7.33566	0.263124	-0.302499	-0.5656242
33	4	42	MIRPJD13	-11.33744	-9.17629	-9.03321	-0.143733	0.178599	0.3225327
34	4	38	SDHRS13	64.639431	55.78784	64.35241	1.435434	1.107370	-0.3280640
35	4	44	KDRPUD13	-37.607225	-37.10497	-39.13793	-0.355237	-0.587052	-0.5308151
36	4	47	ASHGHI13	64.399222	58.27464	62.25918	2.315461	0.474304	-1.8411570
37	4	47	ASHGHI13	64.399222	61.17192	62.25918	-1.787263	-3.628426	-1.8411570
38	4	45	INDGIL13	-22.399722	-19.14042	-20.80954	0.367158	2.457302	1.5901442
39	4	43	SHJ13213	-17.913541	-13.59194	-15.83906	1.447115	2.221595	0.7744789
40	4	43	SHJ13213	-17.913541	-16.56452	-15.03295	-1.625578	-0.851100	0.7744789
41	4	42	SPHGL13	-14.409320	-25.71172	-23.93464	-1.777141	-1.222491	0.5547404
42	4	49	FICHT13	-14.409320	-14.97367	-15.35779	0.539314	-0.404727	-0.9440422
43	4	51	LYLH13	-7.411341	-5.79732	-6.13012	0.422277	1.704034	1.2817383
44	4	52	CHATA13	-7.119933	-9.37467	-7.17477	-1.199913	-1.174736	0.0251770
45	4	53	KSHRS13	-32.229722	-32.13603	-42.17827	0.702247	1.762202	1.7569554
46	4	54	HYVUS13	-37.272675	-33.45537	-38.69609	2.239196	1.522785	-0.7164121
47	4	55	JANLPR13	-13.670963	-13.67521	-14.40229	0.727278	-0.074952	-0.8022308
48	4	57	KHCFN13	-14.421522	-12.97247	-15.80645	-0.766027	-2.580785	-2.3147583
49	4	57	KHCFN13	-14.421522	-19.46935	-19.80645	-0.661903	-2.976661	-2.3147583
50	4	58	MDAPRA13	19.931539	20.06228	19.42547	0.537513	1.131421	0.4939079
51	4	58	MDAPRA13	19.931539	19.77766	19.42547	-0.147819	0.346089	0.4939079

52	4	52	JFSDP113	14.449574	14.25711	13.42551	-1.538401	-1.512461	0.0259399
53	4	50	JHENDAI13	32.260322	22.52303	31.37454	-1.681500	-3.267942	-1.5863419
54	4	62	JHOMZAI13	54.427467	57.06512	54.31004	2.755142	2.665715	-0.0704262
55	4	51	JDTAFL13	33.013247	35.27937	35.02436	-0.314974	2.257029	3.0840034
56	4	52	JDRRAL13	76.015503	74.03313	74.75672	0.266411	-1.135470	-1.2518816
57	4	53	JTHPDI13	27.721741	28.40419	27.68929	0.723922	0.592449	-0.0414729
58	4	53	JSHDPI13	27.721741	26.27202	27.68929	-0.758242	-0.799715	-0.0414729
59	4	73	JAGZFI13	-17.313004	-17.16211	-16.55406	-0.615054	0.143877	0.7589340
60	4	75	JDTGAL13	-5.272345	-6.10027	-5.75546	-0.914811	-1.006811	-0.4220006
61	4	74	JMIGLAE13	-2.039843	-2.32970	-2.42500	-0.135702	-0.680658	-0.4249570
62	4	76	JMSTPDI13	19.026177	18.22515	19.72475	-1.769202	-1.071209	0.5979942
63	4	77	JRIPDI13	30.268546	31.17051	30.16413	1.016366	0.911855	-0.1045108
64	4	77	JRTPDI13	47.227332	48.10583	46.25730	2.148025	2.208488	0.0604630
65	4	67	JHSDRAL13	-34.742833	-35.22303	-35.08462	-0.138419	-0.460201	-0.3217816
66	4	64	JNATPDI13	-74.555664	-73.36723	-75.12253	1.824599	-1.187727	-0.6368627
67	4	57	JHSTAL13	-42.333372	-42.32230	-43.80007	-0.022727	-0.439827	-0.4161000
68	4	55	JAJSHI13	-22.622707	-22.37234	-27.19421	-1.881634	0.619852	2.5014877
69	4	53	JPLSDRI13	-24.742322	-24.41970	-24.63455	0.223837	0.337695	0.1138568
70	4	58	JLSDRI13	-24.742322	-24.44124	-24.63455	-1.314600	-1.700544	0.1138568
71	4	57	JANSDRI13	-22.155334	-21.37163	-20.82482	-0.184738	-0.923752	-0.7390141
72	4	59	JANSDRI13	-22.155334	-22.17002	-23.82482	0.724804	-0.014209	-0.7390141
73	4	70	JSAIDRI13	-12.124003	-13.72952	-17.23203	1.222437	1.325453	0.1359665
74	4	70	JSAIDRI13	-13.124003	-10.76254	-17.23203	-2.331518	-2.645501	0.1359665
75	4	71	JMRDAS13	-25.025122	-25.14374	-25.68817	-0.155103	0.242662	0.3977656
76	4	72	JLSDRI13	-13.030211	-13.77753	-13.03287	-1.737754	-1.777618	-0.0399636
77	4	55	JLSDRI13	-10.722220	-10.20221	-11.85530	0.952769	0.797361	-0.0554085
78	4	80	JHSHI13	-127.573302	-127.55022	-127.78758	0.135661	-0.075625	-0.2122879
79	4	30	JHSHI13	-127.573302	-126.27553	-127.78758	3.412056	3.122768	-0.2122879
80	4	31	JSHDI13	-172.392527	-150.51171	-173.02865	-2.504262	-0.613117	1.8909454
81	4	31	JSHDI13	-172.392527	-175.40728	-173.02865	-1.492302	-1.492302	1.8909454
82	4	26	JKPTAI13	-32.422223	-33.47401	-39.74058	1.266574	0.925904	-1.2406702
83	4	26	JKPTAI13	-32.422223	-32.34554	-38.68023	-0.153601	0.355424	0.5190253
84	4	26	JKPTAI13	-32.422223	-32.33460	-33.34057	1.001965	1.161253	0.1592875
85	4	22	JIKLBA13	-62.722281	-70.31161	-68.77227	-1.539325	-0.511723	1.0276012
86	4	30	JSHDPI13	-18.522224	-15.12244	-17.34239	1.219940	2.470445	1.2505054
87	4	32	JSHDPI13	-21.222402	-24.38255	-22.08142	-1.222133	-2.280252	-0.0818193
88	4	17	JASHGJI13	-52.222243	-51.24382	-54.25269	0.013881	0.051064	0.8371830
89	4	17	JASHGJI13	-52.222243	-53.71562	-52.57434	-1.141285	-0.715774	0.4255116
90	4	17	JASHGJI13	-52.222243	-52.57251	-52.23870	-0.293202	-0.522702	-0.2388000
91	4	17	JASHGJI13	0.002222	0.22437	0.22274	0.723635	0.224379	0.2907429
92	4	78	JADHLL13	-112.222401	-112.03733	-120.25307	1.215741	-0.037869	-1.9536018
93	4	72	JADHLL13	-112.222401	-120.16153	-112.44224	-1.511573	-1.162252	0.3495216
94	4	71	JADHLL13	-112.222401	-112.22442	-112.56907	1.182279	0.712402	-0.4625922
95	4	16	JHSDRI13	-12.422223	-12.15771	-97.73351	-1.718914	-1.957027	-0.2389133
96	4	72	JHSDRI13	-152.622310	-170.32441	-152.25347	-2.055231	-0.638103	1.4178276
97	4	72	JHSDRI13	-152.622310	-125.74152	-137.75486	0.313160	-0.742244	-0.7554054
98	4	43	JHSDRI13	0.000000	0.22252	0.24332	0.722622	0.222504	0.2638835
99	4	51	JYLHFI13	-12.422223	-22.05233	-20.75524	-1.227025	-2.552366	-1.2552795
100	4	36	JYLHFI13	-12.422223	-24.44827	-34.92149	-0.564577	-1.648362	-1.0317642
101	4	36	JYLHFI13	0.000000	-1.17723	-1.64240	0.371467	-1.177237	-1.5424070
102	4	56	JYLHFI13	-12.122215	21.13531	-20.12384	-0.241467	-1.235493	-0.2240267
103	4	75	JYIGAI13	1.522211	0.70373	-2.21764	1.421222	-0.226429	-2.4176531
104	4	22	JYIGAI13	1.322222	0.27547	2.58242	-0.224270	-1.635439	-0.7104706
105	4	70	JYIGAI13	-12.222244	-2.21235	-12.22031	1.077961	1.037611	0.0026500
106	4	57	JANSDRI13	-12.222220	-12.16123	-12.53256	0.327670	-0.161854	-0.5326247

107	4	17	ASHGHI13	-49.912711	-53.75669	-50.03605	0.279357	-0.773972	-1.0533390
108	4	17	ASHGHI13	-49.912711	-53.75669	-50.03605	-3.003423	-4.056763	-1.0533390
109	4	16	GURASLI13	37.266253	36.10713	37.37428	-1.294153	-1.186823	0.1073301
110	4	16	GURASLI13	37.266253	37.20231	37.37428	-0.064981	0.042342	0.1073301
111	4	41	TONGIL13	-127.573571	-127.47439	-126.54446	-0.929927	0.099182	1.0291100
112	4	41	TONGIL13	-127.573571	-126.33291	-126.54446	0.204658	1.233758	1.0291100
113	4	33	ISHODI13	-172.922216	-172.76316	-171.35472	2.611637	1.156044	-1.4555931
114	4	33	ISHODI13	-172.922216	-173.16677	-171.35472	-1.811981	-3.267574	-1.4555931
115	4	26	KAPTALI13	51.792667	50.60264	49.01635	0.474202	-1.301818	-1.7761106
116	4	25	KAPTALI13	51.792667	50.16395	47.19822	0.372563	1.462296	0.4997237
117	4	26	KAPTALI13	71.357727	73.43217	72.56639	0.885790	2.074449	1.1886597
118	4	26	KAPTALI13	71.357727	71.32091	72.56639	-1.225566	-0.036907	1.1886597
119	4	27	CHIRCHI13	22.727173	21.40519	23.85253	-2.357622	-1.292359	1.0650635
120	4	28	MAONHTI13	-6.322700	-7.73266	-6.58502	-1.154839	-1.430165	-0.2753258
121	4	24	MAONHTI13	-6.322700	-7.55619	-6.58502	-0.971156	-1.246481	-0.2753258
122	4	24	MAONHTI13	25.248715	25.27598	25.01735	0.237199	0.726252	-0.2109469
123	4	28	MAONHTI13	25.248725	26.28707	25.03795	1.249218	1.038271	-0.2109468
124	4	28	MAONHTI13	-12.367325	-12.12322	-12.15560	0.242300	0.244092	0.2017856
125	4	29	MAONHTI13	-12.367325	-13.54675	-12.15560	0.519846	0.720632	0.2017856
126	4	30	HATZILI13	31.227745	31.12134	30.93164	0.261694	-0.907402	-0.2691031
127	4	35	FTHILIL13	-9.647872	-9.23399	0.05025	-0.333339	0.384492	0.7178307
128	4	30	HATZILI13	20.215487	20.24508	20.02179	0.153227	0.225616	0.0733197
129	4	22	SKLBAI13	36.110205	39.10791	37.34721	0.760608	1.277524	1.2369156
130	4	22	SKLBAI13	14.881197	17.39433	14.22796	-0.343626	-0.266852	-0.6232262
131	4	30	HATZILI13	25.127745	26.22929	24.93512	1.305794	1.092171	-0.2176230
132	4	30	HATZILI13	25.127746	29.77473	24.93512	-4.210376	-4.423999	-0.2136230
133	4	32	KULSTIL13	3.969133	2.96942	5.03956	-1.220068	-0.090673	1.1294365
134	4	32	KULSTIL13	-15.275432	-16.71584	-15.13423	0.418394	0.560597	0.1421928
135	4	36	COMILAI13	-12.041792	-13.53588	-13.55086	0.224986	-0.674172	-0.7191658
136	4	36	COMILAI13	-12.041702	-13.56285	-13.56086	0.117992	-0.601166	-0.7191658
137	4	36	COMILAI13	2.772930	2.66122	2.22524	0.435286	-0.118702	-0.5539894
138	4	38	SOROSHI13	32.149350	31.26413	31.57733	-0.413703	-0.876212	-0.4630089
139	4	38	SOROSHI13	-47.187219	-47.50722	-47.51634	-1.083963	-1.413088	-0.3291249
140	4	30	SOROSHI13	-47.187219	-46.29443	-47.51634	1.231902	0.922777	-0.3291249
141	4	37	ULLONH13	13.104094	15.51554	18.21746	-1.701813	-2.589349	-0.9865356
142	4	37	ULLONH13	13.104294	17.23318	18.21746	-0.284278	-1.170813	-0.9865356
143	4	37	ULLONH13	-40.003743	-49.12535	-40.52709	0.430744	-0.192493	-0.6232381
144	4	37	ULLONH13	-40.003745	-33.21513	-42.52707	0.711259	0.088721	-0.6232381
145	4	41	TONGIL13	51.205312	51.72772	51.01888	-0.191085	-0.177532	0.0135541
146	4	41	TONGIL13	6.721116	6.39222	7.35778	-0.252421	-0.392918	0.5666733
147	4	37	SOROSHI13	2.456635	19.12329	2.12456	0.272340	0.547270	-0.3320694
148	4	47	PSTALAI13	-2.238952	-62.36351	-60.63057	1.295129	1.592305	0.3051758
149	4	41	TONGIL13	12.122472	12.32161	12.25206	-0.055368	0.476215	0.5315840
150	4	46	GURASLI13	-63.766265	-63.13314	-61.99300	-1.147132	0.636130	1.7832632
151	4	46	GURASLI13	-63.766265	-61.41227	-61.99300	0.572136	2.353399	1.7832632
152	4	44	KAPALI13	22.407399	21.85596	20.92328	1.861877	-0.751441	-1.6133184
153	4	47	ASHGHI13	15.202607	14.26350	15.11442	-0.844227	-1.623385	-0.7783890
154	4	47	ASHGHI13	15.202607	15.77989	15.11442	0.564325	-0.113993	-0.7783890
155	4	43	SHIJZHI13	24.676921	22.22322	24.06334	-1.779054	-2.328515	-0.5584598
156	4	42	SHIJZHI13	14.422341	15.28927	15.44580	0.534248	1.490211	-0.2559631
157	4	53	PHONSHI13	7.424968	6.58649	6.13922	0.418263	-0.857575	-1.2858391
158	4	31	GYLITTI13	7.211124	5.22722	7.13522	-1.192757	-1.214265	-0.0245094
159	4	47	ASHGHI13	51.207822	50.03663	50.03430	0.002349	-1.792359	-1.7947073
160	4	53	KAPALI13	37.625214	41.62642	42.37454	2.251940	3.091254	0.7493138
161	4	34	NYHOSHI13	12.672272	15.22652	14.42717	0.712428	1.926294	0.8074760

152	4	56	GILPRA13	18.510010	18.32634	18.82324	-0.738871	1.574360	2.3132324
163	4	56	GILPRA13	18.510010	18.55014	18.82324	-0.173091	2.140140	2.3132324
164	4	57	KHCHN113	-18.674227	-20.22523	-17.15339	-1.573531	-2.151925	-0.4784644
165	4	57	KHCHN113	-18.674227	-16.21949	-19.15332	2.334904	1.855437	-0.4784644
166	4	58	NDAPRA13	-16.152700	-17.73249	-16.16212	-1.550352	-1.579737	-0.0293851
167	4	58	NDAPRA13	-31.202432	-31.42410	-30.40513	-1.018977	0.485336	1.5043135
168	4	59	JHSDF113	-51.459702	-62.72946	-51.40170	2.602249	2.662756	0.0675082
169	4	60	JHSDF113	-51.459702	-55.17732	-54.83589	-0.342005	-3.316913	-2.9748077
170	4	61	BTAIL13	-74.210042	-73.32647	-73.62432	0.367838	1.583576	1.2157373
171	4	62	BTAIL13	-74.210042	-75.22642	-74.21787	0.893843	0.932455	0.0386119
172	4	62	BTAIL13	-74.210042	-77.74531	-76.81787	-0.227462	-0.888030	0.0386119
173	4	64	CDLPRAL3	17.419457	15.75022	16.65048	-0.892559	-1.567540	-0.7679820
174	4	73	BAGRHT13	2.312157	4.21552	3.80854	-0.922949	-0.395562	0.4963875
175	4	73	BAGRHT13	2.001122	2.31172	2.42681	-0.185106	0.310518	0.4958245
176	4	73	BRIGAL13	-17.735217	-21.55378	-17.41117	-2.142607	-2.847962	-0.7052541
177	4	75	MRIOR13	-22.526448	-22.93912	-22.53343	0.427308	0.557295	-0.0572774
178	4	77	FRIDOR13	-43.265572	-42.33500	-44.00329	1.168274	1.133561	-0.0347137
179	4	64	NATDRE13	35.549464	34.87331	35.87359	-1.052274	-0.717151	0.3331244
180	4	63	ISADRI13	75.426373	74.17108	77.07277	1.071313	1.744222	0.6522869
181	4	63	ISADRI13	50.642723	52.74502	51.22324	1.651185	2.095222	0.4440367
182	4	64	NATDRE13	32.115204	25.65271	27.54734	-1.487637	-4.455494	-2.5678568
183	4	67	BDZAL13	25.073502	24.23596	24.95555	-0.718575	-0.935533	-0.1179576
184	4	67	BDZAL13	25.073502	24.06991	24.95555	-0.896529	-1.004487	-0.1172576
185	4	68	PLSBR113	20.322202	22.76345	21.14543	-0.381276	0.365232	0.7472157
186	4	68	PLSBR113	20.322202	22.07730	21.14543	0.928372	1.675537	0.7472157
187	4	69	RAVSPR13	13.255227	10.72189	19.06649	0.655324	0.465899	-0.1894951
188	4	69	RAVSPR13	13.255227	17.79113	18.05649	-2.285367	-2.474862	-0.1894951
189	4	70	SALDRI13	26.242574	25.32156	25.83703	-0.155462	-0.566005	-0.4105449
190	4	71	PJRSAS13	13.036233	11.32577	13.12160	-1.724932	-1.689260	0.0355721
191	4	63	ISADRI13	11.739120	11.77346	12.73149	0.841760	0.321542	0.0497818
192	4	70	ASRILL23	122.714233	123.36445	122.73851	3.825046	4.247584	0.2215385
193	4	78	ASRILL23	122.714233	123.35502	122.73851	-1.383425	-1.161956	0.2215385
194	4	72	CHASL23	122.622475	125.50323	127.42636	-1.792416	-4.104518	-2.1121025
195	4	72	CHASL23	122.622475	126.36337	127.42636	-1.132488	-3.244590	-2.1121025
196	4	1	KPTIEX60	72.422202	71.023504	72.74959	1.225256	2.535226	1.2406702
197	4	2	KPTIEX60	39.122263	39.52517	39.68923	-0.155770	-0.574726	-0.5190253
198	4	2	KPTIEX60	31.422262	31.35775	31.24257	1.027156	0.957872	-0.1522875
199	4	4	SKLSTX11	67.722251	67.17562	67.77227	-1.575653	-2.424255	-1.0276012
200	4	5	STSTX10	12.522223	12.61051	12.24223	1.261227	0.010722	-1.2505054
201	4	6	STSTX11	21.022203	22.22235	22.09142	-2.107074	-2.027255	0.0818193
202	4	7	ASTNEX84	52.722245	54.22771	54.25249	0.714025	-0.923159	-0.8371830
203	4	8	ASTNEX84	52.222247	51.42213	52.57434	-1.152223	-1.575712	-0.4255116
204	4	9	ASTNEX84	51.222201	52.22235	51.23372	-0.285257	-0.044467	0.2389090
205	4	10	ASTNEX84	2.810200	2.41326	-0.229074	0.704711	0.413268	-0.2907429
206	4	11	ASTNEX84	113.222221	122.22220	122.25307	1.250221	3.004533	1.2536018
207	4	12	ASTNEX84	113.222221	117.41262	117.64224	-1.532225	-1.892847	-0.3495216
208	4	13	ASTNEX84	118.222243	120.52222	119.46227	1.203223	1.673412	0.4625892
209	4	14	ASTNEX84	37.422203	35.28222	37.73891	-1.754724	-1.515811	0.2389133
210	4	16	STSIJ210	134.522210	135.15304	132.25347	-2.115440	-3.533268	-1.4179276
211	4	16	STSIJ210	134.222245	135.76350	135.75436	0.013637	0.759243	0.7534054
212	4	17	STSIJ210	3.022202	2.46624	-0.26338	0.730950	0.466267	-0.2638835
213	4	18	STSIJ210	12.422202	12.44322	22.75724	-1.311422	-0.055212	1.2552725
214	4	19	KALIK110	32.222225	33.42246	34.08142	-0.584042	0.497722	1.0817642
215	4	20	KALIK110	3.022202	1.22210	1.54240	0.378629	1.922105	1.5494070
216	4	21	KALIK110	12.122214	12.21222	20.19334	-0.958144	0.035882	1.2940267

217	4	20	OSL2X77	-1.500011	2.37092	2.31764	1.457182	3.870835	2.4176531
218	4	23	YRAXX20	-1.350271	-1.55742	-0.58747	-7.957929	-0.257452	0.7104706
219	4	24	SRRSLY20	10.222265	17.12584	10.29031	1.135527	1.125877	-0.0096500
220	4	25	MSPIX20	18.222233	12.25742	12.57056	0.418055	0.257677	0.5326247
221	4	28	AS HULL23	11.112717	48.25575	52.03605	-1.080286	-0.026247	1.0533390
222	4	28	AS HULL23	42.222712	42.45544	51.03605	-0.481580	0.471758	1.0533390
223	4	29	GRASL23	-37.272223	-37.57442	-37.37428	-0.220191	-0.327521	-0.1073301
224	4	29	GRASL23	-37.272223	-37.57442	-37.37428	-1.153557	-1.260888	-0.1073301
225	4	30	TONGIL27	127.572272	121.43384	126.54446	-3.110592	-4.139709	-1.0291100
226	4	30	TONGIL27	127.572272	127.42574	126.54446	1.081275	0.252166	-1.0291100
227	4	31	IS HULL23	172.822216	132.22492	181.35479	2.532136	3.987782	1.4555931
228	4	31	IS HULL23	172.822216	172.41247	181.35479	-1.735305	-0.279712	1.4555931
229	5	27	CHIR213	-22.372237	-2.10201	-22.85753	22.257534	22.372436	-0.5851090
230	5	28	HADV123	-14.274227	-14.00225	-14.14338	1.140444	1.283985	0.1435399
231	5	30	HATZL13	-12.155233	-11.43411	-11.38563	-0.248673	0.731571	0.7801950
232	5	30	HATZL13	-12.155233	-12.23322	-11.38563	-1.842644	-1.059450	0.7801950
233	5	28	HADV123	-14.274227	-13.12763	-14.20576	1.009123	1.336455	0.3273308
234	5	29	SYKLAB13	7.272217	2.22225	8.44250	1.055354	1.925141	0.8697867
235	5	27	SYKLAB13	7.272217	2.25545	8.44250	-0.537046	0.282740	0.8697867
236	5	32	KULSIL13	-11.416516	-1.32222	-10.80442	1.315104	2.127128	0.8120239
237	5	32	KULSIL13	-11.416516	-2.25571	-10.40442	1.548778	2.363802	0.8120239
238	5	30	HATZL13	-1.152772	-1.35422	-1.47227	-0.385653	-0.702211	-0.3165580
239	5	30	HATZL13	-1.152772	0.14111	-1.47227	1.622324	1.303235	-0.3165580
240	5	35	FENILL13	-2.522272	-2.05215	-2.05133	2.892175	2.541613	-0.3575623
241	5	35	COHILA13	3.222222	2.53222	2.61552	0.058380	-0.595442	-0.6648302
242	5	35	COHILA13	-4.522231	-6.22215	-7.04219	0.114035	-0.328932	-0.4428685
243	5	33	HLGHA213	-22.732272	-23.72222	-23.82246	0.107383	-1.056486	-1.1638699
244	5	31	DRHP213	-7.122255	-7.14022	-7.07583	-0.244184	0.052932	0.1041174
245	5	34	BPALIA13	-2.611511	-2.57222	-2.78834	0.212752	0.035917	-0.1768351
246	5	34	BPALIA13	-2.611511	-10.22228	-2.78834	-0.442508	-0.625343	-0.1768351
247	5	33	HLGHA213	3.222271	2.33227	2.60078	-0.051508	-0.726224	-0.7347863
248	5	34	BPALIA13	2.221682	2.32724	1.08444	1.243428	0.106169	-1.1372385
249	5	38	SDIP213	2.222242	2.10122	4.48458	0.618705	1.152421	0.5437169
250	5	38	SDIP213	2.222242	2.31422	4.48458	0.822416	1.373132	0.5437169
251	5	37	CHIR213	-4.622223	-6.22224	-5.03753	-1.388213	-2.225750	-0.3375370
252	5	32	HULL2013	-40.011527	-12.40252	-38.42833	0.222744	1.505223	1.5132484
253	5	36	GHASL13	21.011222	22.55222	21.53355	2.021270	2.545582	0.5243123
254	5	36	GHASL13	21.011222	24.57222	21.53355	3.032508	3.563820	0.5243123
255	5	40	DIAN213	-2.224612	-2.76522	-2.10333	0.337535	0.468814	0.1312722
256	5	42	JAN213	-7.224612	-2.44222	-2.17333	-0.344355	-0.213587	0.1312722
257	5	41	TONGIL13	-2.222222	-5.70222	-2.75424	-2.248114	-3.007083	-0.0509691
258	5	41	TONGIL13	-2.222222	-1.45222	-2.75424	1.305222	1.245317	-0.0509691
259	5	42	HIR213	12.222162	-11.12222	-10.81211	-0.379102	1.104732	1.4840422
260	5	43	PSTGL13	4.612223	4.25122	4.25268	-0.210807	-0.351041	-0.3502343
261	5	42	HIR213	-17.224221	-17.24777	-16.45803	-0.582746	0.247085	0.8368313
262	5	39	SAIR213	32.124227	32.15422	34.42243	-1.052454	-2.825284	-1.7458291
263	5	44	KOR213	-12.124612	-16.10251	-16.10771	0.087201	1.066407	0.7748203
264	5	47	ADIS213	-4.572248	-6.74727	-4.65752	-2.022378	-2.175104	-0.0847250
265	5	47	ADIS213	-4.572248	-4.11222	-4.65752	0.532328	0.454583	-0.0847250
266	5	45	ISGAIL13	-12.222116	-12.22156	-11.08135	2.557724	0.378555	-0.1812323
267	5	43	SAJI1313	3.022222	2.21473	4.32877	-0.494043	0.785531	1.2805738
268	5	48	SHJI1313	2.022222	1.22222	4.32877	-0.711621	0.362383	1.2905738
269	5	49	IRHUSL13	-2.224612	-2.22457	-2.13253	0.145053	-0.529711	-0.7747650
270	5	52	FNCH213	-5.221752	-7.36522	-6.22422	-0.572458	-1.703625	-1.1331673
271	5	51	SYLH1313	-6.022223	-6.22221	-3.17522	-1.764414	-0.912256	0.8461587

272	5	52	CHATAK13	-3.400000	-2.100000	-4.36326	2.152723	1.290220	-0.8689821
273	5	53	KSHR5113	-21.354117	-25.18972	-23.72877	-1.360022	-3.224479	-1.8544571
274	5	54	MY4NS513	-16.200734	-16.47286	-16.85020	0.377351	-0.252109	-0.6294608
275	5	55	JAHUP113	-5.100000	-5.200000	-5.63254	-0.665565	0.301981	0.2675469
276	5	57	KHCENT13	-50.000000	-50.15327	-49.35490	-0.808388	-0.114470	0.6939173
277	5	57	KHCENT13	-50.000000	-50.53273	-49.35490	-1.177823	-0.483906	0.6939173
278	5	59	INDPRA13	-31.200734	-31.20000	-31.37335	-3.452570	-4.525255	-1.0725851
279	5	58	INDPRA13	-31.200734	-31.65951	-31.37335	0.713849	-0.358736	-1.0725851
280	5	59	JG303713	-34.310000	-36.65473	-35.13319	0.479458	0.657522	0.1790643
281	5	59	JAHUP113	-5.100000	-5.200000	-5.63254	-3.665387	-3.121625	0.5437613
282	5	52	GIRPRA13	-14.400000	-14.27326	-14.71056	-2.263414	-2.123933	0.1394808
283	5	51	BHATA113	-14.200000	-13.45825	-15.19072	-2.118241	-1.272355	0.9459853
284	5	52	GIRPRA13	-14.400000	-7.16778	-5.81001	-0.355856	-2.481354	-2.4249878
285	5	53	ISHD1113	-23.620000	-12.00733	-33.95752	1.950192	1.815289	-0.1449108
286	5	57	ISHD1113	-23.620000	-33.60182	-33.26752	-0.634282	-0.779173	-0.1449108
287	5	73	BAJUP113	-2.450000	-1.23000	-1.65908	-0.373121	-1.481503	-1.1093022
288	5	75	BRISAL13	1.000000	1.53400	1.92680	0.007197	0.529723	0.5217258
289	5	74	MANGLA13	-1.100000	-1.43882	-0.58260	-0.855218	-0.438801	0.4174169
290	5	75	INDP1113	-27.640000	-27.62000	-21.74862	-3.673649	-2.772604	0.9010434
291	5	77	FRIDP113	-17.670000	-16.98514	-15.51125	-0.174974	1.770502	1.9453764
292	5	52	GIRPRA13	-14.400000	-14.40525	-12.54653	-1.358717	-3.557973	-1.5091615
293	5	57	BHATA113	-14.200000	2.66700	3.37651	-0.314512	-0.649543	0.1659691
294	5	64	MATP1113	-14.600000	-13.02357	-13.21631	-2.007204	-1.303410	0.7037033
295	5	57	BHATA113	-2.150000	-1.12712	-1.72344	0.555353	1.032854	0.3664907
296	5	59	BAJUP113	-1.400000	-1.40000	-13.65174	-0.355852	-0.108603	0.7491694
297	5	58	PLESR113	7.301357	6.27000	7.38500	-0.406944	-0.413292	-0.0063479
298	5	58	PLESR113	7.301357	6.27000	7.38500	-0.533430	-0.639779	-0.0063479
299	5	59	RANGP113	1.021358	2.36703	7.30431	2.062719	1.345692	-0.7170379
300	5	69	RANGP113	1.021358	6.27000	7.30431	-1.025878	-1.742917	-0.7170379
301	5	70	SAJUP113	2.280000	2.43673	2.06329	0.373454	0.146964	-0.2264894
302	5	70	SAJUP113	2.280000	0.21700	2.06329	-1.846217	-2.072707	-0.2264894
303	5	71	MURPAG13	-11.100000	-10.23197	-9.11753	-1.114236	0.893312	2.0075493
304	5	72	THAKUP13	-6.200000	-4.71701	-5.31486	0.596945	1.582053	0.9851090
305	5	55	GJUP1113	5.000000	-2.48565	-3.64890	0.963252	2.514351	1.5510998
306	5	50	INDP1113	-31.200734	-37.22343	-39.17553	1.180003	0.403271	-0.7768035
307	5	50	INDP1113	-31.200734	-31.20000	-39.17553	5.778401	5.001593	-0.7768035
308	5	41	ISHD1113	-23.620000	-23.00161	-5.21986	5.829255	5.612885	-0.2153690
309	5	81	ISHD1113	-23.620000	-8.41636	-5.21986	-2.496499	-2.711868	-0.2153690
310	5	26	KAPTAL13	-21.250000	-21.40000	-22.23569	-0.554317	0.353866	0.9081841
311	5	26	KAPTAL13	-21.250000	-12.66627	-13.85187	-1.607078	-1.780765	-0.1756668
312	5	23	KAPTAL13	-21.250000	-21.20000	-27.70371	-0.887012	-0.551110	0.3359020
313	5	22	GKIBAL13	-27.400000	-4.40670	-43.33575	-0.160033	0.081485	1.1424236
314	5	48	SHR1113	-24.400000	-22.20000	-41.40000	3.462674	2.667873	-0.7946014
315	5	39	SHR1113	-24.400000	-5.76626	-51.42627	-5.273567	-5.214869	-0.6442904
316	5	47	ASHD1113	-14.310000	-2.55263	-2.33275	-0.247385	0.753779	1.0036640
317	5	47	ASHD1113	-14.310000	-2.20000	-2.20000	-2.257622	-2.757405	-0.4207930
318	5	47	ASHD1113	-2.171000	-3.79113	-2.25162	-1.329512	-1.234033	0.0254792
319	5	47	ASHD1113	-2.730000	-6.98577	-7.12634	0.510555	1.046311	0.5357563
320	5	72	ASHD1113	-21.200000	-29.70000	-21.13024	0.400405	1.096875	0.8964807
321	5	78	ASHD1113	-21.200000	-27.82400	-29.67378	-2.152305	-0.292361	1.1529446
322	5	78	ASHD1113	-21.200000	-29.12000	-20.35762	0.157280	1.636385	1.4621048
323	5	46	SHR1113	-24.400000	-23.20000	-25.44212	-2.434118	0.126319	0.6304383
324	5	79	SHR1113	-24.400000	-48.12000	-50.69570	1.807093	1.392143	-0.4149497
325	5	79	SHR1113	-24.400000	-53.21174	-52.43357	-1.478157	-1.416033	0.0612259
326	5	48	SHR1113	-14.400000	-14.31236	-14.46334	-0.667121	1.103788	1.7729101

382	5	47	AFICMUI3	-5.420781	-6.26753	-6.72165	0.634121	-0.664985	-1.2991066
383	5	48	SHUJ3E13	6.963262	8.32529	7.71939	0.375510	1.132845	0.7563353
384	5	49	SRMNGE13	3.566581	4.43322	4.74105	-0.307750	0.866748	1.1745081
385	5	50	FUCM4E13	2.466416	-0.30721	1.57222	-1.504510	-2.466701	-0.8521916
386	5	51	SYLIL113	1.212308	4.13225	2.73418	2.149065	3.012944	0.8718781
387	5	52	ASICMUI3	22.810229	23.14173	24.57664	-1.434950	0.328964	1.7638140
388	5	53	KSCM2E13	15.214277	17.12011	16.67263	0.327533	1.153899	0.7663667
389	5	54	NYMNGE13	4.421463	2.22052	3.42538	-0.664853	-1.600133	-0.9352807
390	5	55	QULORAI3	49.218753	49.22427	48.53299	-0.532615	-1.211762	-0.6791472
391	5	56	QULORAI3	49.218753	47.18114	48.53299	-1.395430	-2.067577	-0.6791472
392	5	57	KACM1E13	33.013916	33.14756	34.12572	-0.272149	0.133563	1.1128120
393	5	57	KACM1E13	33.013916	33.46744	34.12572	-0.572272	0.433540	1.1128120
394	5	57	MDAPAI3	35.213622	35.70293	35.89203	0.375037	0.683432	-0.1916051
395	5	58	MDAPAI3	24.428571	24.46767	25.83702	-1.159335	-1.960890	-0.7915556
396	5	59	JT503E13	14.313217	17.26287	18.22086	-0.147298	-0.370341	-0.2223432
397	5	60	JITMDAI3	17.211121	16.75967	17.35031	-0.300115	-0.437433	-0.5373180
398	5	61	QITALE13	7.537164	7.21113	7.92522	1.106229	3.425283	-2.3183584
399	5	62	H4MMAI3	35.832154	35.52523	35.98762	-0.311620	-0.182217	0.1224732
400	5	62	H4MMAI3	35.832154	30.12586	35.29762	2.442243	2.577721	0.1224732
401	5	66	QULORAI3	-1.127612	-0.16672	-0.10732	-0.059307	1.015242	1.2752563
402	5	77	BAG3HT13	-3.235207	-4.22813	-4.45007	0.414242	-0.092221	-0.5141538
403	5	73	BAG3HT13	-3.411510	-1.58291	-0.82293	-0.353322	-1.271307	-0.4172250
404	5	75	BPISAL13	21.174309	17.67541	20.24018	-2.544772	-3.479286	-0.9343147
405	5	76	NRIP1E13	17.257224	15.62320	15.43643	1.129774	-0.212195	-2.1102695
406	5	77	FR100E13	11.356556	14.32025	13.52416	0.725027	2.463700	1.7376118
407	5	64	HAT32E13	-3.756237	-2.55592	-1.22248	1.345539	1.700293	-0.1462460
408	5	63	IS120E13	12.401325	17.12223	17.31951	-0.182112	-0.852609	-0.6634891
409	5	63	IS120E13	1.211234	1.74001	2.27743	-2.235815	-2.540617	-0.3032019
410	5	64	HAT32E13	14.022417	12.22567	13.11771	-0.231240	-1.941238	-0.9608984
411	5	67	Q533AI13	-2.523223	-2.45922	-2.53276	-0.211146	-0.927323	-0.0166774
412	5	67	AR30AI13	-0.223123	-0.26537	-0.53974	0.274384	0.257705	-0.0166774
413	5	68	PL503E13	-2.422123	-7.22167	-3.75680	0.775123	1.508432	0.7333040
414	5	68	PL503E13	-2.422123	-2.07482	-8.75680	0.581918	1.415222	0.7333040
415	5	69	RAN50E13	-3.432248	-3.24447	-3.32663	-1.547775	-1.434531	0.2131637
416	5	69	RAN50E13	-3.432248	-3.22723	-3.32668	0.168745	0.381907	0.2131637
417	5	70	SA120E13	12.751057	7.57705	8.72362	-1.125557	-3.173988	-2.0474310
418	5	71	PJ30AS13	4.022232	6.41746	3.21825	2.592512	-0.402363	-1.0079735
419	5	63	ISHRJI13	2.322216	1.52350	0.73550	0.358001	-0.630313	-1.5883150
420	5	78	AS10L23	36.722245	34.53121	37.55871	-0.223921	-0.112030	0.8048713
421	5	78	AS10L23	36.722245	22.55325	37.55871	1.220258	2.725122	0.8049713
422	5	79	GHRASL23	22.024226	22.00222	24.17221	-1.262227	-2.125741	-0.8638144
423	5	79	GHRASL23	22.024226	21.32227	24.17221	-2.343141	-1.206955	-0.9638144
424	5	1	KPT12X40	21.122414	26.86322	28.45414	-1.592177	-2.324479	-0.7342815
425	5	2	KPT12X50	14.402227	15.22225	14.42222	0.535441	0.522407	0.0222654
426	5	3	KPT12X50	36.322222	33.25522	36.02222	-2.042224	-2.421438	-0.3785133
427	5	4	SKL12X30	33.021227	32.55222	32.52222	0.023114	-1.357705	-1.3738213
428	5	5	S3C12X10	51.245122	51.54215	42.77491	2.357345	3.626765	0.9226122
429	5	6	S3C12X33	62.513511	61.31653	63.24485	-1.282221	-1.126290	0.7312417
430	5	7	ASS12X34	12.735324	12.14712	12.61421	-0.465025	-1.522410	-1.1225157
431	5	8	ASS11E13	19.112144	19.35222	18.54953	0.121518	0.536205	0.4353881
432	5	9	ASS11E32	5.074224	7.21222	5.02220	2.885958	0.836247	-0.0508199
433	5	10	ASS12J62	7.222244	6.22222	7.22224	-0.351083	-0.892106	-0.5480230
434	5	11	ASS12J50	31.322221	22.22222	31.42232	-2.345670	-2.753944	-0.4172742
435	5	12	ASS22J150	11.242221	22.22222	30.57704	-2.288824	-3.552376	-1.2535527
436	5	13	ASS32J150	31.242221	22.22222	30.32230	-2.158321	-3.613672	-1.4553003

437	5	14	GR5LX20	32.22473	32.22473	32.22473	1.164784	0.536070	-0.6298946
438	5	15	GR5LX20	61.50534	61.50534	61.50534	0.404667	0.633251	0.2285839
439	5	16	GR5LX20	14.00361	14.00361	14.00361	-1.715720	-1.589040	0.0276804
440	5	17	SHJ375T	14.00361	14.00361	14.00361	-0.726091	-2.542393	-1.8163023
441	5	18	SYL11X20	21.53829	21.53829	21.53829	0.393575	0.629245	0.2356709
442	5	19	KIL11X20	39.27397	39.27397	39.27397	-1.033977	-2.559339	-1.5253420
443	5	20	KIL11X20	39.27397	39.27397	39.27397	0.133734	-1.274459	-1.4081255
444	5	21	KIL11X20	21.53829	21.53829	21.53829	1.105206	0.938728	-0.1662791
445	5	22	BRSLX20	32.22473	32.22473	32.22473	2.783508	0.300455	-2.4831533
446	5	23	BRSLX20	53.57966	53.57966	53.57966	-0.379514	-1.042596	-0.6639712
447	5	24	BRSLX20	29.53150	29.53150	29.53150	0.070387	-0.120767	-0.1913548
448	5	25	RHSR1X20	24.43677	24.43677	24.43677	-3.203499	-3.177004	0.1064837
449	5	78	ASJUL23	-3.10410	-3.10410	-3.10410	2.718874	3.827350	0.9047557
450	5	78	ASJUL23	-2.42642	-2.42642	-2.42642	0.672617	1.584372	0.9047557
451	5	79	BRAS123	25.48461	25.48461	25.48461	3.023270	2.159570	-0.2647203
452	5	79	BRAS123	25.48461	25.48461	25.48461	0.267577	-0.517122	-0.8547203
453	5	80	TONG123	38.24029	38.24029	38.24029	-0.355428	-0.222337	-0.1569097
454	5	80	TONG123	38.24029	38.24029	38.24029	-3.253379	-3.410309	-0.1569092
455	5	81	15JUL23	4.55673	4.55673	4.55673	-1.323972	-2.473353	-1.1493827
456	5	81	15JUL23	4.55673	4.55673	4.55673	-0.995490	-2.144870	-1.1493807
457	1	1	KPT11X30	1.000000	1.000000	1.000000	0.000000	0.000000	0.0000000
458	1	2	KPT11X30	1.000000	1.000000	1.000000	-0.000384	0.000000	0.0003920
459	1	3	KPT12X30	1.000000	1.000000	1.000000	0.000268	0.000061	-0.0002962
460	1	4	SKLSTX11	1.000000	1.000000	1.000000	-0.000053	-0.000701	-0.0006485
461	1	5	SDSHX33	1.000000	1.000000	1.000000	0.000265	0.000182	0.0009174
462	1	6	SDSHX33	1.000000	1.000000	1.000000	-0.000762	-0.000139	0.0006227
463	1	7	AS5H2X34	1.000000	1.000000	1.000000	0.000052	-0.001001	-0.0010647
464	1	8	AS5H1167	1.000000	1.000000	1.000000	-0.000480	0.000362	0.0008430
465	1	9	AS5H1X30	1.000000	1.000000	1.000000	-0.000293	-0.000474	-0.0001812
466	1	10	AS5H2150	1.000000	1.000000	1.000000	0.000322	-0.000500	-0.0009230
467	1	11	AS5H1150	1.000000	1.000000	1.000000	0.000037	0.001259	0.0005722
468	1	12	AS5H1150	1.000000	1.000000	1.000000	0.000015	0.000145	0.0001297
469	1	13	AS5H1150	1.000000	1.000000	1.000000	0.000579	0.000535	-0.0000422
470	1	14	GR5LX20	1.000000	1.000000	1.000000	-0.000437	-0.000447	-0.0000114
471	1	15	GR5LX20	1.000000	1.000000	1.000000	0.001611	0.001611	0.0000000
472	1	16	GR5LX20	1.000000	1.000000	1.000000	-0.000052	-0.000092	-0.0001450
473	1	17	SHJ375T	1.000000	1.000000	1.000000	0.000032	0.000332	0.0003004
474	1	18	SYL11X20	1.000000	1.000000	1.000000	0.000131	0.001193	0.0010614
475	1	19	KIL11X20	1.000000	1.000000	1.000000	-0.000337	-0.000241	0.0001564
476	1	20	KIL11X20	1.000000	1.000000	1.000000	-0.000277	-0.000573	-0.0004152
477	1	21	KIL11X20	1.000000	1.000000	1.000000	-0.000353	0.000743	0.0010977
478	1	22	BRSLX20	1.000000	1.000000	1.000000	-0.000232	-0.002454	-0.0022221
479	1	23	BRSLX20	1.000000	1.000000	1.000000	0.000472	0.001227	0.0007357
480	1	24	BRSLX20	1.000000	1.000000	1.000000	0.000003	0.000023	0.0000247
481	1	25	RHSR1X20	1.000000	1.000000	1.000000	0.000174	0.001297	0.0011225

THE MEASUREMENTS MARKED WITH ASTERISKS ARE SIMULATED AS BAD DATA

VALUES OF PERFORMANCE INDICES

PERFORMANCE INDEX AS A FUNCTION OF: COMPUTED VALUE P.U. THRESHOLD VALUE P.U.

1. (ISP.-EST.) 2.2320236 1.2500070

2. (EST.-TRUE) 0.33029487 0.46421130

