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RASP-IDENT

Subspace Model Identification Programs

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CONTENTS

1. Introduction 1
2. The State Intersection Class of Subspace Model Identification 3
 - RPIMN4 - N4SID Deterministic-Stochastic Identification 4
3. The MOESP Class of Subspace Model Identification 13
 - RPIMOE - Ordinary MOESP Identification 16
 - RPIMPI - Ordinary MOESP Identification with Past Inputs 22
 - RPIMPO - Ordinary MOESP Identification with Past Inputs and Outputs 28
 - RPIMRS - Ordinary MOESP Identification with Reconstructed States and Past Inputs 34
 - RPIMKG - MOESP-SI Deterministic-Stochastic Identification 40
4. Auxiliary Tools for System Identification 49
 - IMMPID - Estimation of Markov Parameters from Input-Output Data 50
 - IMSHFT - I/O Data Shifting to Compensate Input Vector Dead-Times 57
 - IMICID - Estimation of the Initial Conditions 59
 - IMCLID - Closed-Loop Identification of Linear Systems 63
 - RPIMIU - Simulation of Linear Time-Invariant Discrete-Time Systems 68
- Appendix 1. RASP-IDENT Driver Routines 73
- Appendix 2. SLICOT Compatible Driver Routines 75
- Appendix 3. SLICOT Compatible Computational and Auxiliary Routines 76
- Appendix 4. Called RASP-MODRED, LAPACK and BLAS Routines 78

1. Introduction

Subspace Model Identification (SMI) is a recently developed promising branch in system identification. One feature that makes this approach so appealing is that SMI formulates the identification of a linear time-invariant (LTI) system as a column space approximation problem followed by a set of linear least-squares problems. As a consequence, instead of using iterative optimisation schemes as are required in the classical parametric model identification approach documented e.g. in /1/, SMI schemes use basic numerical linear algebra tools, such as QR factorization and SVD. The attractiveness of SMI techniques is further increased by the limited number of parameters (essentially only one) the user has to select to determine the model structure and this without any restriction on model generality. For a further discussion of the features of SMI we refer to the specialized literature, such as /2/, /3/.

The RASP-IDENT package includes implementations of two families of SMI techniques to identify an LTI system. The first one is referred to as the State Intersection (SI) class of SMI techniques. The implemented SI method is described in /2/. The second one is referred to as the Multivariable Output Error State Space (MOESP) class of techniques. The implemented MOESP methods are described in /3/, /5/ and /6/.

In addition to the identification facilities, auxiliary routines are available that allow to detect dead-time and to shift appropriately the input-output signals, to estimate the initial state, to identify the system while operating in closed-loop, to simulate discrete-time LTI systems.

Literature:

/1/ Ljung L.

System Identification Toolbox.
The Math Works, Inc., 1992.

/2/ Van Overschee P. and De Moor B.

N4SID: Two subspace algorithms for the identification of combined deterministic-stochastic systems.
Automatica, Vol.30, No.1, pp.75-93, 1994.

/3/ Verhaegen M.

Identification of the deterministic part of MIMO state space models given in innovations form from input-output data.
Automatica, Vol.30, No.1, pp.61-74, 1994.

- /4/ Moonen M., De Moor B., Vandenberghe L. and Vandewalle J.
On- and off-line identification of linear state-space models.
Int. J. Control, Vol. 49, pp. 219-232, 1989.
- /5/ Verhaegen M. and Dewilde P.
Subspace Model Identification. Part 1: The output-error state-
space model identification class of algorithms.
Int. J. Control, 56, pp. 1187-1210, 1992.
- /6/ Verhaegen M.
Subspace Model Identification. Part 3: Analysis of the
ordinary output-error state-space model identification
algorithm.
Int. J. Control, 58, pp. 555-586, 1993.

2. The State Intersection Class of Subspace Model Identification

The main feature of the State Intersection class of SMI techniques is the determination of the state sequence of the LTI system to be identified or of an observer to reconstruct its state sequence via the intersection of the row spaces of matrices constructed from "past" and "future" input-output data. The basic idea was first described in /1/. An extension of this idea led to the N4SID algorithm developed in /2/. This algorithm allows to identify LTI state space models in so-called innovation form. This form considers the additive perturbations acting on the output of the system to be modelled in the following way:

$$\begin{aligned}x_{k+1} &= Ax_k + Bu_k + Kv_k \\y_k &= Cx_k + Du_k + v_k\end{aligned}$$

where v_k is a zero-mean white noise sequence and u_k is deterministic input sequence (perfectly known to the user), that is statistically independent from the user. Both the basic and the extended variant of this class produce statistically consistent and efficient estimates when the stipulated assumptions hold.

The following subroutine is available for the SI class of SMI:

RPIMN4 computes, by using the basic N4SID algorithm, consistent and statistically efficient estimates of the matrices (A,B,C,D) of a state space model when v is generated by an innovation model.

/1/ Moonen M., De Moor B., Vandenberghe L. and Vandewalle J.
On- and off-line identification of linear state-space models.
Int. J. Control, Vol. 49, pp. 219-232, 1989.

/2/ Van Overschee P. and De Moor B.
N4SID: Two subspace algorithms for the identification of combined deterministic-stochastic systems.
Automatica, Vol.30, No.1, pp.75-93, 1994.

SUBROUTINE RPIMN4

N4SID Deterministic-Stochastic Identification

Procedure purpose:

This subroutine determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices and, optionally, the noise covariance matrices and the Kalman gain matrix using the "subspace state space system identification" (N4SID) algorithm. It is assumed that the input is a zero-mean white noise process.

The model structure is :

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) + w(k) \\y(k) &= Cx(k) + Du(k) + e(k)\end{aligned}$$

where $w(k)$ and $e(k)$ are zero-mean white noise sequences, independent of the input $u(k)$. Matrix D could be zero.

The Kalman predictor structure is :

$$\begin{aligned}\bar{x}(k+1) &= A\bar{x}(k) + Bu(k) + K(y(k) - C\bar{x}(k) - Du(k)) \\ \bar{y}(k) &= C\bar{x}(k) + Du(k)\end{aligned}$$

where K is the Kalman gain matrix stored in FK .

There is an option for sequential input-output data processing.

Usage:

```
CALL RPIMN4( U, NSMP, M, Y, L, NOBL, TOL, WITHD, WITHCV, WITHK, CONCT,
            LAST, CTRL, N, A, B, C, D, FK, R, IWORK, RWORK, LRWORK, *)
```

```
U      : IN, DOUBLE (NSMP,M)
        system input data sequence matrix U = [u_1 u_2 ... u_m].
        Column j of U contains the NSMP values of the
        j-th input component for consecutive time increments.
NSMP   : IN, INTEGER
        number of rows of matrices U and Y (number of samples)
        (When sequential data processing is used, NSMP is the
        number of samples of the current data batch.)
        NSMP >= 2*(M+L+1)*NOBL - 1, for non-sequential processing
        of the data;
        NSMP >= 2*NOBL, for sequential processing of several data
        batches;
        NSMP = 0, and LAST = .TRUE., is allowed if the upper
```

triangular factor of the QR factorization is already available in the array R, and only the system matrices are to be computed.

The total number of samples when calling with LAST = .TRUE. should be at least $2*(M+L+1)*NOBL - 1$.

- M : IN, INTEGER
dimension of system input vector.
- Y : IN, DOUBLE (NSMP,L)
system output data sequence matrix $Y = [y_1 \ y_2 \ \dots \ y_l]$.
Column j of Y contains the NSMP values of the j-th output component for consecutive time increments.
- L : IN, INTEGER
dimension of system output vector.
- NOBL : IN, INTEGER
number of block rows in the processed input and output Hankel matrices. NOBL should be chosen larger than n, the estimated dimension of state vector.
- TOL : IN, DOUBLE
absolute tolerance used for determining an estimate of the system order. If $TOL > 0$, the estimate is indicated by the number of singular values greater than or equal to TOL. (Singular values less than TOL are considered zero.)
If $TOL = 0$, the internally computed default value $TOL = NOBL*EPS*SV[1]$ is used, where EPS is the relative machine precision, and SV[1] is the maximal singular value.
If $TOL < 0$, the estimate is indicated by the index of the singular value that has the largest logarithmic gap to its successor. (See Method for the definition of the singular values.)
- WITHD : IN, LOGICAL
specifies whether or not a non-zero feedthrough matrix D to be included in the estimated state space representation:
WITHD = .TRUE. means a possibly non-zero D matrix should be estimated.
WITHD = .FALSE. means a zero D matrix is assumed.
- WITHCV: IN, LOGICAL
specifies whether or not the covariance matrices are to be computed:
WITHCV = .TRUE. means the covariance matrices are computed.
WITHCV = .FALSE. means the covariance matrices should not be computed.
The covariance matrices are automatically computed if
WITHK = .TRUE..
- WITHK : IN, LOGICAL
specifies whether or not the Kalman gain matrix K is to be computed:
WITHK = .TRUE. means the Kalman gain matrix is computed.

WITHK = .FALSE. means the Kalman gain matrix is not computed.

CONCT : IN, LOGICAL
 specifies whether or not the successive data batches belong to a single experiment:
 CONCT = .TRUE. means the current data batch is a continuation of the previous data batch.
 CONCT = .FALSE. means there is no connection between the current data batch and the previous ones.

LAST : IN, LOGICAL
 specifies whether or not sequential data processing is used, and (for sequential processing) whether or not the current data block is the last one, as follows:
 LAST = .TRUE. means non-sequential processing or the last data block in sequential processing.
 LAST = .FALSE. means not the last data block in sequential processing. In this case the only calculation performed is to update the triangular factor of the QR factorization.

CTRL : IN, LOGICAL
 specifies whether or not the user's confirmation of the system order estimate is desired before the computation of system matrices, as follows:
 CTRL = .TRUE. means user's confirmation is desired.
 CTRL = .FALSE. means no confirmation is necessary.
 If CTRL = .TRUE., a reverse communication routine, IMCORD, is called, and, after inspecting the singular values and system order estimate, n , the user may accept n or set a new value.
 IMCORD is not called by the routine if CTRL = .FALSE..
 CTRL = .TRUE. should be used in extreme cases only.

N : OUT, INTEGER
 estimated order of the system.

A : OUT, DOUBLE (NOBL-1,NOBL-1)
 the $N \times N$ estimated system state matrix A.
 (column dense)

B : OUT, DOUBLE (NOBL-1,M)
 the $N \times M$ estimated system input matrix B.
 (column dense)

C : OUT, DOUBLE (L,NOBL-1)
 the $L \times N$ estimated system output matrix C.
 (column dense)

D : OUT, DOUBLE (L,M)
 the $L \times M$ estimated system feedthrough matrix D if WITHD = .TRUE..
 D is not referenced if WITHD = .FALSE..
 (column dense)

FK : OUT, DOUBLE (NOBL-1,L)
 if WITHK = .TRUE., on normal return, or when the warning errors 1 or 3 are issued, the leading N*L part of this array contains the Kalman gain matrix K. FK is not referenced when WITHK = .FALSE..
 (column dense)

R : IN, OUT, DOUBLE (2*(M+L)*NOBL,2*(M+L)*NOBL)
 On output, if LAST = .FALSE. then R contains the current upper triangular factor of the QR factorization used to compress the data before determining the singular values in the N4SID algorithm.
 The content of R should be preserved between successive calls of RPIMN4 with LAST = .FALSE..
 On first input, the content of R is not meaningful.
 During the computations with LAST = .TRUE., R contains some relevant data for the N4SID algorithm.
 On output with LAST = .TRUE., the content of R is destroyed; if WITHCV = .TRUE. or WITHK = .TRUE., on normal return, and in the absence of warning error 2, the first N+L rows and columns of R contain the estimated covariance matrix of the vector [w', e']';
 if WITHK = .TRUE., on normal return, or when the warning errors 1 or 3 are issued, the rows N+L+1 to 2*N+L+1 and the columns 1 to N of R contain the estimation error covariance matrix X (the Riccati equation solution).

IWORK : OUT, INTEGER((M+L)*NOBL)
 working array.

RWORK : OUT, DOUBLE (LRWORK)
 working array.
 On exit, if LAST = .TRUE., and normal return, then:
 RWORK(1) contains the optimal suggested value of LRWORK;
 RWORK(2) contains an estimate of the reciprocal condition number for the linear algebraic system solved for computing the system matrices (in the absence of warning error 2);
 RWORK(3:L*NOBL+2) contain singular values used for determining the system order.
 If WITHK = .TRUE., LAST = .TRUE., on normal return, or when the warning errors 1 or 3 are issued, then:
 RWORK(L*NOBL+3) contains an estimate of the reciprocal condition number of the computed output covariance matrix Ry;
 RWORK(L*NOBL+4) contains an estimate of the reciprocal condition number of the computed matrix A;
 RWORK(L*NOBL+5) contains an estimate of the reciprocal condition number of the computed matrix U_11, used to obtain the Riccati equation solution X;
 RWORK(L*NOBL+6) contains an estimate of the reciprocal condition number of the matrix Ry + C*X*C', used to

obtain the Kalman gain matrix;
 RWORK(L*NOBL+7:L*NOBL+N+6) contain the real parts of the optimal (Kalman predictor) poles;
 RWORK(L*NOBL+N+7:L*NOBL+2*N+6) contain the imaginary parts of the optimal (Kalman predictor) poles.
 Part of this information is also available for warning errors 4 to 9, if logically possible.
 The first (M+L)*(2*NOBL-1) elements of RWORK should be preserved between successive calls of the routine with LAST = .FALSE., and the final call with LAST = .TRUE., if successive data batches belong to a single experiment (CONCT = .TRUE.).

On return with error number 4, RWORK(1) contains the minimum necessary value of LRWORK.

LRWORK: IN, INTEGER

dimension of working array RWORK.

The value of LRWORK must be at least LRWMIN, where

$$\begin{aligned} \text{LRWMIN} = \max(& 4*(\text{NOBL}+1)*(M+L)*\text{NOBL}, 5*(M+L)*\text{NOBL}, \\ & 2+L*\text{NOBL} + \\ & \max(2*(\text{NOBL}-1) + \max((2*M+L)*\text{NOBL}+L, \\ & (L*\text{NOBL}-L)*(\text{NOBL}-1)), 4*(\text{NOBL}-1)+L, \\ & 4*(M*\text{NOBL}+\text{NOBL}-1), \\ & 4*(L*\text{NOBL}+M), 5*(L*\text{NOBL}+M)-4, \\ & 3 + \max(10*(\text{NOBL}-1), 3*L), \\ & 5 + 2*(\text{NOBL}-1) + \max(\text{NOBL}-1, 3*L))) \end{aligned}$$

For NSMP = 0, the first term of the first max expression could be omitted. For WITHK = .FALSE., the last two terms of the second max expression could be omitted.

This estimate is computed using the largest value for n, namely NOBL-1.

For good performance, LRWORK should be larger.

* : RETURN 1, target label in case of error (e.g. *1111)

File input/ output:

none

Method:

Let us denote $m = M$, $l = L$, $t = \text{NSPM}$, $s = \text{NOBL}$.

1) For non-sequential data processing, the $t \times 2(m+1)s$ matrix

$$\begin{bmatrix} U' & Y' \\ 1,2s,t & 1,2s,t \end{bmatrix}$$

is constructed, where U and Y are Hankel matrices $1,2s,t$ $1,2s,t$ defined in terms of the input and output data [1].

A QR factorization is used to compress the data, an oblique projection is computed in terms of some submatrices of the upper triangular factor R [2], and then a singular value decomposition (SVD) of the submatrix

$$R_{22} := R(ms+1:(2m+1)s, ms+1:(2m+1)s)$$

of R reveals the order n of the system as the number of "non-zero" singular values. System matrices are finally computed using the first n right singular vectors of R_{22} , and other submatrices of R , solving a linear algebraic system in a total least squares sense [3]. The covariance matrices are computed by solving several least squares problems, constructed based on the residuals.

2) For sequential data processing, the QR decomposition is done sequentially, by updating the upper triangular factor R . When all data have been compressed, the system order and system matrices are computed as in the previous case.

The Kalman gain is obtained by solving a discrete-time algebraic matrix Riccati equation using the Schur vectors approach [4] for the dual of an optimal control problem.

Literature

- /1/ Van Overschee, P., and De Moor, B.
N4SID: Two Subspace Algorithms for the Identification of Combined Deterministic-Stochastic Systems.
Automatica, Vol.30, pp. 75-93, 1994.
- /2/ Van Overschee, P.
Subspace Identification : Theory - Implementation - Applications.
Ph. D. Thesis, Katholieke Universiteit Leuven, 1995.
- /3/ Van Huffel, S., and Vandewalle J.
The Total Least Squares Problem: Computational Aspects and Analysis.
SIAM, Philadelphia, 1991.
- /4/ Laub A. J.
A Schur method for solving algebraic Riccati equations.
IEEE Trans. Automat. Control, AC-24, pp. 913-921, 1979.

Remarks:

- The NSMP argument may vary from a cycle to another in sequential data processing, but NOBL, M, and L should be kept constant. For efficiency, it is advisable to use NSMP as large as possible.
- When 100 cycles of sequential data processing are completed with LAST = .FALSE., a warning is issued, to prevent for an infinite loop.

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Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

Consider the discrete-time system

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) + Ew(k) \\y(k) &= Cx(k) + Du(k) + Fv(k)\end{aligned}$$

with the following matrices:

$$A = \begin{pmatrix} 1.5 & -0.7 \\ 1.0 & 0.0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0.5 \end{pmatrix} \quad D = 0$$

$$E = \begin{pmatrix} 0.0005 & 0.0 \\ 0.0 & 0.006 \end{pmatrix} \quad F = \begin{pmatrix} 0.0528 \end{pmatrix}$$

whose output response $y(t)$ to normal (0,1) random input signals $u(t)$, $w(t)$, $v(t)$ and zero initial state is determined by using the subroutine RPIMIU.

The following sequence of statements can be used to estimate the order and the matrices of the original system

```

DATA   ISEED/ 1, 2, 3, 5 /
NSMP   = 120
NOBL   = 6
M      = 1
MW     = 2
L      = 1
LV     = 1
TOL    = -1.000
WITHD  = .FALSE.
WITHK  = .TRUE.
LAST   = .TRUE.
CONCT  = .TRUE.
CTRL   = .TRUE.
LRWORK = 8100
C      Compute the output response for uniformly distributed
C      input sequence.
X(1)   = 0.000
X(2)   = 0.000
CALL RPIMIU( NSMP, 2, M, L, MW, LV, A, B, E, C, D, F, X, U,
*          W, V, ISEED, Y, RWORK, LRWORK, WITHD, 'G',
*          'G', 'R', 'R', 'R', 3, *1111 )

```

```

C   Perform the system identification.
    CALL RPIMN4( U, NSMP, M, Y, L, NOBL, TOL, WITHD, WITHCV,
*           WITHK, CONCT, LAST, CTRL, N, A, B, C, D, FK,
*           R, IWORK, RWORK, LRWORK, *1111 )

```

The order detection singular values are:

```

.87524260D+02  .42090540D+02  .32603442D+00  .22092251D+00  .19614963D+00
.16624421D+00

```

We can take the system order $n = 2$.

The estimated system matrices are:

$$\bar{A} = \begin{pmatrix} .8245 & -.5418 \\ .2643 & .6748 \end{pmatrix} \quad \bar{B} = \begin{pmatrix} 3.8116 \\ -1.6200 \end{pmatrix} \quad \bar{C} = \begin{pmatrix} .5093 & .5822 \end{pmatrix}$$

The estimated Kalman gain matrix K is:

$$K = \begin{pmatrix} .1190 \\ .1635 \end{pmatrix}$$

The estimated (inter-)covariance matrix (for $[Ew; Fv]$) is:

$$\begin{pmatrix} .0001 & .0002 & .0003 \\ .0002 & .0006 & .0000 \\ .0003 & .0000 & .0029 \end{pmatrix}$$

The estimated state covariance matrix X is:

$$\begin{pmatrix} .0012 & -.0001 \\ -.0001 & .0011 \end{pmatrix}$$

The relative output error is: $\|y(t) - \bar{y}(t)\| / \|y(t)\| = .41315D-02$

Error Messages:

-1-

Invalid parameter value on entry.

-2-

Singular value decomposition failed.

-3-

A singular upper triangular matrix was found.

-4-

Not enough working storage. It should be at least //LENG//.

Warnings Messages:

-1-

More than 100 cycles in sequential data processing

-2-

All singular values are exactly zero (N = 0)

-3-

The least squares problem for covariances is rank-deficient

-4-

The computed output covariance matrix Ry is too small'

-5-

The covariance matrix Ry is numerically singular

-6-

The transformed matrix A is numerically singular

-7-

Schur form computation did not converged

-8-

Schur form ordering failed

-9-

The computed matrix U_11 is numerically singular

3. The MOESP Class of Subspace Model Identification

The main feature of this class of SMI techniques is the determination of the extended observability matrix of the deterministic part of the MOESP model. When this model is given in the following way:

$$\begin{aligned}x_{k+1} &= Ax_k + Bu_k + w_k \\y_k &= Cx_k + Du_k + v_k\end{aligned}$$

the extended observability matrix Γ_s is given as:

$$\Gamma_s = \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{s-1} \end{bmatrix}$$

The basic idea was first described in /1,2/. The algorithm developed from this idea, which is indicated by the ordinary MOESP (OM) scheme, allows to identify in a statistically consistent and efficient way LTI systems that can be described by the MOESP model with $w_k = 0$ and v_k being zero-mean white noise independent from the input.

Later on, an extension was formulated based on the incorporation of instrumental variables based on past input quantities and/or reconstructed state variables that allow to consistently identify MOESP class of models for v_k being a zero-mean arbitrary stochastic disturbance independent from the input, and $w_k = 0$. The price paid for this increased applicability is that the estimates are no longer efficient. This variant is indicated as the PI scheme (when past inputs are used as instrumental variables) and the RS scheme (when reconstructed state variables are used).

When the additive disturbance w_k in the MOESP model is generated by an innovation model of the form $w_k = Kv_k$ and v_k is a zero-mean white noise independent of the input, it is again possible to obtain both consistent and efficient estimates when, in addition to the past input, past output quantities are used as instrumental variables. The scheme derived from this extension of ordinary MOESP scheme is indicated as the PO scheme.

For the deterministic-stochastic identification, it is possible to estimate the deterministic part by using the MOESP approach with the PO scheme, and the stochastic part, that is the noise covariance matrices, by using the Space-Intersection (SI) approach /4/. Further, the corresponding Kalman gain K can be computed to allow the design of state observers /3/.

The algorithms belonging to the MOESP class have the striking feature of being highly streamlined, in the sense that the sequence of

computations performed by these schemes is almost independent from the type of problems to be analyzed. This allows to modularize the implementations of the algorithms within this class to a great extent.

The following routines are available for MOESP class of SMI:

- RPIMOE computes, by using the ordinary MOESP scheme, consistent and statistically efficient estimates of the matrices (A,B,C,D) of a state space model when $w = 0$ and v is zero-mean white noise independent from the input.
- RPIMPI computes, by using the ordinary MOESP scheme extended with instrumental variables based on past input quantities, consistent estimates of the matrices (A,B,C,D) of a state space model when $w = 0$ and v is zero-mean but of arbitrary statistical color.
- RPIMPO computes, by using the ordinary MOESP scheme extended with instrumental variables based on past input and past output quantities, consistent and statistically efficient estimates of the matrices (A,B,C,D) of a state space model when w is generated by an innovation model and v is zero-mean white noise independent from the input.
- RPIMRS computes, by using the ordinary MOESP scheme extended with instrumental variables based on reconstructed state and/or past input quantities, consistent estimates of the matrices (A,B,C,D) of a state space model when $w = 0$ and v is zero-mean but of arbitrary statistical color.
- RPIMKG in addition to the RPIMPO routine, also estimates by using the SI approach the noise covariance matrices and the Kalman gain to allow the design of a state observer as a state predictor.

Literature:

- /1/ Verhaegen M. and Dewilde P.
Subspace Model Identification. Part 1: The output-error state-space model identification class of algorithms.
Int. J. Control, 56, pp. 1187-1210, 1992.
- /2/ Verhaegen M.
Subspace Model Identification. Part 3: Analysis of the ordinary output-error state-space model identification algorithm.
Int. J. Control, 58, pp. 555-586, 1993.

/3/ Verhaegen M.

Identification of the deterministic part of MIMO state space models given in innovations form from input-output data. Automatica, Vol.30, No.1, pp.61-74, 1994.

/4/ Van Overschee P. and De Moor B.

N4SID: Two subspace algorithms for the identification of combined deterministic-stochastic systems. Automatica, Vol.30, No.1, pp.75-93, 1994.

SUBROUTINE RPIMOE

Ordinary MOESP identification

Procedure purpose:

This subroutine estimates from given input and output data sequences, the order and the matrices of the linear time-invariant discrete-time state space model

$$x(k+1) = Ax(k) + Bu(k)$$

$$y(k) = Cx(k) + Du(k) + v(k)$$

by using the ordinary MOESP algorithm.

Usage:

```
CALL RPIMOE(U, NSMP, M, Y, L, NOBL, TOL, WITHD, CONCT, LAST,
           CTRL, N, A, B, C, D, R, IWORK, RWORK, LRWORK, *)
```

U : IN, DOUBLE (NSMP,M)
 system input data sequence matrix $U = [u_1 \ u_2 \ \dots \ u_m]$.
 Column j of U contains the NSMP values of the
 j -th input component for consecutive time increments.

NSMP : IN, INTEGER
 number of rows of matrices U and Y (number of samples)
 (When sequential data processing is used, NSMP is the
 number of samples of the current data batch.)
 NSMP $\geq (M+L+1)*NOBL - 1$, for non-sequential processing
 of the data;
 NSMP $\geq NOBL$, for sequential processing of several data
 batches;
 NSMP = 0, and LAST = .TRUE., is allowed if the upper
 triangular factor of the QR factorization is
 already available in the array R , and only the
 system matrices are to be computed.
 The total number of samples when calling with LAST = .TRUE.
 should be at least $(M+L+1)*NOBL - 1$.

M : IN, INTEGER
 dimension of system input vector.

Y : IN, DOUBLE (NSMP,L)
 system output data sequence matrix $Y = [y_1 \ y_2 \ \dots \ y_l]$.
 Column j of Y contains the NSMP values of the
 j -th output component for consecutive time increments.

L : IN, INTEGER
 dimension of system output vector.

NOBL : IN, INTEGER
number of block rows in the processed input and output
Hankel matrices. NOBL should be chosen larger than n ,
the estimated dimension of state vector.

TOL : IN, DOUBLE
absolute tolerance used for determining an estimate of the
system order. If $TOL > 0$, the estimate is indicated by the
number of singular values greater than or equal to TOL.
(Singular values less than TOL are considered zero.)
If $TOL = 0$, the internally computed default value
 $TOL = NOBL * EPS * SV[1]$ is used, where EPS is the relative
machine precision, and SV[1] is the maximal singular value.
If $TOL < 0$, the estimate is indicated by the index
of the singular value that has the largest logarithmic
gap to its successor. (See Method for the definition of
the singular values.)

WITHD : IN, LOGICAL
specifies whether or not a non-zero feedthrough matrix D
to be included in the estimated state space representation:
WITHD = .TRUE. means a possibly non-zero D matrix should be
estimated.
WITHD = .FALSE. means a zero D matrix is assumed.

CONCT : IN, LOGICAL
specifies whether or not the successive data batches belong
to a single experiment:
CONCT = .TRUE. means the current data batch is a
continuation of the previous data batch.
CONCT = .FALSE. means there is no connection between the
current data batch and the previous ones.

LAST : IN, LOGICAL
specifies whether or not sequential data processing is
used, and (for sequential processing) whether or not the
current data block is the last one, as follows:
LAST = .TRUE. means non-sequential processing or the last
data block in sequential processing.
LAST = .FALSE. means not the last data block in sequential
processing. In this case the only calculation
performed is to update the triangular factor
of the QR factorization.

CTRL : IN, LOGICAL
specifies whether or not the user's confirmation of the
system order estimate is desired before the computation
of system matrices, as follows:
CTRL = .TRUE. means user's confirmation is desired.
CTRL = .FALSE. means no confirmation is necessary.
If CTRL = .TRUE., a reverse communication routine,
IMCORD, is called, and, after inspecting the singular

values and system order estimate, n , the user may accept n or set a new value.
 IMCORD is not called by the routine if CTRL = .FALSE..
 CTRL = .TRUE. should be used in extreme cases only.

N : OUT, INTEGER
 estimated order of the system.

A : OUT, DOUBLE (NOBL-1,NOBL-1)
 the N*N estimated system state matrix A.
 (column dense)

B : OUT, DOUBLE (NOBL-1,M)
 the N*M estimated system input matrix B.
 (column dense)

C : OUT, DOUBLE (L,NOBL-1)
 the L*N estimated system output matrix C.
 (column dense)

D : OUT, DOUBLE (L,M)
 the L*M estimated system feedthrough matrix D if
 WITHD = .TRUE..
 D is not referenced if WITHD = .FALSE..
 (column dense)

R : IN, OUT, DOUBLE ((max(M,L)+L)*NOBL,(M+L)*NOBL)
 On output, if LAST = .FALSE., then the leading
 (M+L)*NOBL x (M+L)*NOBL part of R contains the current
 upper triangular factor of the QR factorization used to
 compress the data before determining the singular values
 in the MOESP algorithm.
 The content of R should be preserved between successive
 calls of RPIMOE with LAST = .FALSE..
 On first input, the content of R is not meaningful.

IWORK : OUT, INTEGER(M)
 working array.

RWORK : OUT, DOUBLE (LRWORK)
 working array.
 RWORK(1) contains the optimal suggested value of LRWORK;
 RWORK(2) contains an estimate of the reciprocal condition
 number for the final linear algebraic system solved;
 RWORK(3:2+L*NOBL) contain the singular values used for
 determining the system order.
 The first (M+L)*(NOBL-1) elements of RWORK should
 be preserved between successive calls of the routine with
 LAST = .FALSE., and the final call with LAST = .TRUE.,
 if successive data batches belong to a single experiment
 (CONCT = .TRUE.).
 On return with error number 4, RWORK(1) contains the
 minimum necessary value of LRWORK.

LRWORK: IN, INTEGER
 dimension of working array RWORK.

The value of LRWORK must be at least LRWMIN, where

$$\text{LRWMIN} = \max((\text{NOBL}+2)*(\text{M}+\text{L})*\text{NOBL},$$

$$2+\text{L}*\text{NOBL} + \max(((2*\text{NOBL}-3)*\text{L}+2)*(\text{NOBL}-1),$$

$$(\text{L}*\text{NOBL}-1)*\text{M}*\text{NOBL},$$

$$4*(\text{L}*\text{NOBL}+\text{M}),$$

$$5*(\text{L}*\text{NOBL}+\text{M})-4))$$

For NSMP = 0, the first term of the first max expression could be omitted.

This estimate is computed using the largest and smallest values for n, namely NOBL-1 and 1, respectively (for positive and negative terms, respectively).

In practice, n could be much smaller than NOBL-1.

For good performance, LRWORK should be larger.

* : RETURN 1, target label in case of error (e.g. *1111)

File input/ output:

none

Method:

Let us denote $m = M$, $l = L$, $t = \text{NSPM}$, $s = \text{NOBL}$.

For non-sequential data processing, the $t \times (m+1)s$ matrix

$$\begin{bmatrix} U' & Y' \\ 1,s,t & 1,s,t \end{bmatrix}$$

is constructed, where U and Y are Hankel matrices
 $1,s,t$ $1,s,t$

defined in terms of the input and output data /1/.

A QR factorization is used to compress the data, and then a singular value decomposition (SVD) of the principal submatrix $R_{22} := R(ms+1:(m+1)s, ms+1:(m+1)s)$ of the upper triangular factor R reveals the order n of the system as the number of "non-zero" singular values. System matrices are finally computed from the right singular vectors of R_{22} and the submatrices R_{11} and R_{12} of R .

For sequential data processing, the QR decomposition is done sequentially, by updating the upper triangular factor R . When all data have been compressed, the system order and system matrices are computed as in the previous case.

Literature

/1/ Verhaegen M., and P. Dewilde

Subspace Model Identification. Part 1: The output-error state-space model identification class of algorithms.

Int. J. Control, 56, pp. 1187-1210, 1992.

/2/ Verhaegen M.

Subspace Model Identification. Part 3: Analysis of the ordinary output-error state-space model identification algorithm.

Int. J. Control, 58, pp. 555-586, 1993.

Remarks:

- The NSMP argument may vary from a cycle to another in sequential data processing, but NOBL, M, and L should be kept constant. For efficiency, it is advisable to use NSMP as large as possible.
- When 100 cycles of sequential data processing are completed with LAST = .FALSE., a warning is issued, to prevent for an infinite loop.

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Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

Consider the discrete-time system (A,B,C,D) with the following matrices:

$$A = \begin{pmatrix} 1.5 & -0.7 \\ 1.0 & 0.0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0.5 \end{pmatrix} \quad D = 0$$

whose output response $y(t)$ to random input signals $u(t)$ and zero initial state is determined by using the subroutine RPIMIU.

The following sequence of statements can be used to estimate the order and the matrices of the original system

```

DATA   ISEED/ 1, 2, 3, 5 /
NSMP   = 120
NOBL   = 10
M      = 1
L      = 1
TOL    = -1.0D0
WITHD  = .FALSE.
LAST   = .TRUE.
CONCT  = .TRUE.
CTRL   = .TRUE.
LRWORK = 8100
C      Compute the output response for uniformly distributed
C      input sequence.

```

```

X(1)  = 0.0D0
X(2)  = 0.0D0
CALL RPIMIU( NSMP, 2, M, L, 0, 0, A, B, E, C, D, F, X, U,
*           W, V, ISEED, Y, RWORK, LRWORK, WITHD, 'G',
*           'G', 'R', 'R', 'R', 1, *1111 )
C      Perform the system identification.
CALL RPIMOE( U, NSMP, M, Y, L, NOBL, TOL, WITHD, CONCT, LAST,
*           CTRL, N, A, B, C, D, R, IWORK, RWORK, LRWORK,
*           *1111 )

```

The order detection singular values are:

```
3.2203D+01  1.4612D+01  0  0  0  0  0  0  0  0
```

The estimated system matrices are:

$$\bar{A} = \begin{pmatrix} .7831 & .5606 \\ -.2472 & .7169 \end{pmatrix} \quad \bar{B} = \begin{pmatrix} -3.5468 \\ -2.3707 \end{pmatrix} \quad \bar{C} = \begin{pmatrix} -.5749 & .4382 \end{pmatrix}$$

which coincide with the matrices obtained by applying a similarity transformation to the original system.

The relative output error is: $\|y(t) - \bar{y}(t)\| / \|y(t)\| = 0.11277D-14$

Error Messages:

```

-1-
Invalid parameter value on entry.
-2-
Singular value decomposition failed.
-3-
A singular upper triangular matrix was found.
-4-
Not enough working storage. It should be at least //LENG//.

```

Warnings Messages:

```

-1-
More than 100 cycles in sequential data processing
-2-
All singular values are exactly zero (N = 0)

```

SUBROUTINE RPIMPI

Ordinary MOESP identification with past inputs

Procedure purpose:

This subroutine estimates from given input and output data sequences, the order and the matrices of the linear time-invariant discrete-time state space model

$$x(k+1) = Ax(k) + Bu(k)$$

$$y(k) = Cx(k) + Du(k) + v(k)$$

by using the ordinary MOESP algorithm with past inputs.

Usage:

```
CALL RPIMPI(U, NSMP, M, Y, L, NOBL, TOL, WITHD, CONCT, LAST,
           CTRL, N, A, B, C, D, R, IWORK, RWORK, LRWORK, *)
```

U : IN, DOUBLE (NSMP,M)
system input data sequence matrix $U = [u_1 \ u_2 \ \dots \ u_m]$.
Column j of U contains the NSMP values of the j -th input component for consecutive time increments.

NSMP : IN, INTEGER
number of rows of matrices U and Y (number of samples).
(When sequential data processing is used, NSMP is the number of samples of the current data batch.)
NSMP $\geq (2*M+L+2)*NOBL - 1$, for non-sequential processing of the data;
NSMP $\geq 2*NOBL$, for sequential processing of several data batches;
NSMP = 0, and LAST = .TRUE., is allowed if the upper triangular factor of the QR factorization is already available in the array R, and only the system matrices are to be computed.
The total number of samples when calling with LAST = .TRUE. should be at least $(2*M+L+2)*NOBL - 1$.

M : IN, INTEGER
dimension of system input vector.

Y : IN, DOUBLE (NSMP,L)
system output data sequence matrix $Y = [y_1 \ y_2 \ \dots \ y_l]$.
Column j of Y contains the NSMP values of the j -th output component for consecutive time increments.

L : IN, INTEGER
dimension of system output vector.

NOBL : IN, INTEGER
number of block rows in the processed input and output
Hankel matrices. NOBL should be chosen larger than n ,
the estimated dimension of state vector.

TOL : IN, DOUBLE
absolute tolerance used for determining an estimate of the
system order. If $TOL > 0$, the estimate is indicated by the
number of singular values greater than or equal to TOL.
(Singular values less than TOL are considered zero.)
If $TOL = 0$, the internally computed default value
 $TOL = NOBL * EPS * SV[1]$ is used, where EPS is the relative
machine precision, and SV[1] is the maximal singular value.
If $TOL < 0$, the estimate is indicated by the index
of the singular value that has the largest logarithmic
gap to its successor. (See Method for the definition of
the singular values.)

WITHD : IN, LOGICAL
specifies whether or not a non-zero feedthrough matrix D
to be included in the estimated state space representation:
WITHD = .TRUE. means a possibly non-zero D matrix should be
estimated.
WITHD = .FALSE. means a zero D matrix is assumed.

CONCT : IN, LOGICAL
specifies whether or not the successive data batches belong
to a single experiment:
CONCT = .TRUE. means the current data batch is a
continuation of the previous data batch.
CONCT = .FALSE. means there is no connection between the
current data batch and the previous ones.

LAST : IN, LOGICAL
specifies whether or not sequential data processing is
used, and (for sequential processing) whether or not the
current data block is the last one, as follows:
LAST = .TRUE. means non-sequential processing or the last
data block in sequential processing.
LAST = .FALSE. means not the last data block in sequential
processing. In this case the only calculation
performed is to update the triangular factor
of the QR factorization.

CTRL : IN, LOGICAL
specifies whether or not the user's confirmation of the
system order estimate is desired before the computation
of system matrices, as follows:
CTRL = .TRUE. means user's confirmation is desired.
CTRL = .FALSE. means no confirmation is necessary.
If CTRL = .TRUE., a reverse communication routine,
IMCORD, is called, and, after inspecting the singular

values and system order estimate, n , the user may accept n or set a new value.

IMCORD is not called by the routine if CTRL = .FALSE..
 CTRL = .TRUE. should be used in extreme cases only.

N : OUT, INTEGER
 estimated order of the system.

A : OUT, DOUBLE (NOBL-1,NOBL-1)
 the N*N estimated system state matrix A.
 (column dense)

B : OUT, DOUBLE (NOBL-1,M)
 the N*M estimated system input matrix B.
 (column dense)

C : OUT, DOUBLE (L,NOBL-1)
 the L*N estimated system output matrix C.
 (column dense)

D : OUT, DOUBLE (L,M)
 the L*M estimated system feedthrough matrix D if
 WITHD = .TRUE..
 D is not referenced if WITHD = .FALSE..
 (column dense)

R : IN, OUT, DOUBLE ((max(2*M,L)+L)*NOBL,(2*M+L)*NOBL)
 On output, if LAST = .FALSE., then the leading
 (2*M+L)*NOBL x (2*M+L)*NOBL part of R contains the current
 upper triangular factor of the QR factorization used to
 compress the data before determining the singular values in
 the MOESP algorithm.
 The content of R should be preserved between successive
 calls of RPIMPI with LAST = .FALSE..
 On first input, the content of R is not meaningful.

IWORK : OUT, INTEGER(M)
 working array.

RWORK : OUT, DOUBLE (LRWORK)
 working array.
 RWORK(1) contains the optimal suggested value of LRWORK;
 RWORK(2) contains an estimate of the reciprocal condition
 number for the final linear algebraic system solved;
 RWORK(3:2+min(M,L)*NOBL) contain the singular values used
 for determining the system order.
 The first $M*(2*NOBL-1) + L*(NOBL-1)$ elements of RWORK
 should be preserved between successive calls of the routine
 with LAST = .FALSE., and the final call with LAST = .TRUE.,
 if successive data batches belong to a single experiment
 (CONCT = .TRUE.).
 On return with error number 4, RWORK(1) contains the
 minimum necessary value of LRWORK.

LRWORK: IN, INTEGER
 dimension of working array RWORK.

The value of LRWORK must be at least LRWMIN, where

$$\text{LRWMIN} = \max(2*(\text{NOBL}+1)*(2*M+L)*\text{NOBL}, 2+\min(M,L)*\text{NOBL} +$$

$$\max(((2*\text{NOBL}-3)*L+2)*(\text{NOBL}-1),$$

$$(L*\text{NOBL}-1)*M*\text{NOBL}, 4*(L*\text{NOBL}+M),$$

$$5*(L*\text{NOBL}+M)-4))$$

For NSMP = 0, the first term of the first max expression could be omitted.

This estimate is computed using the largest and smallest values for n, namely NOBL-1 and 1, respectively (for positive and negative terms, respectively).

In practice, n could be much smaller than NOBL-1.

For good performance, LRWORK should be larger.

* : RETURN 1, target label in case of error (e.g. *1111)

File input/ output:

none

Method:

Let us denote $m = M$, $l = L$, $t = \text{NSPM}$, $s = \text{NOBL}$.

For non-sequential data processing, the $t \times (2m+1)s$ matrix

$$\begin{bmatrix} \text{Uf}' & \text{Up}' & \text{Yf}' \\ s+1,s,t & 1,s,t & s+1,s,t \end{bmatrix}$$

is constructed, where Up , Uf and Yf are $1,s,t$, $s+1,s,t$ and $s+1,s,t$

Hankel matrices defined in terms of the input and output data /2/.

A QR factorization is used to compress the data, and then a singular value decomposition (SVD) of the submatrix

$R_{23} := R(ms+1:2ms, 2ms+1:(2m+1)s)$ of the upper triangular factor

R reveals the order n of the system as the number of "non-zero" singular values. System matrices are finally computed from the right singular vectors of R_{23} and the submatrices R_{11} and R_{13} of R .

For sequential data processing, the QR decomposition is done sequentially, by updating the upper triangular factor R . When all data have been compressed, the system order and system matrices are computed as in the previous case.

Literature

/1/ Verhaegen M., and P. Dewilde

Subspace Model Identification. Part 1: The output-error state-space model identification class of algorithms.

Int. J. Control, 56, pp. 1187-1210, 1992.

/2/ Verhaegen M.

Subspace Model Identification. Part 3: Analysis of the ordinary output-error state-space model identification algorithm.

Int. J. Control, 58, pp. 555-586, 1993.

Remarks:

- The NSMP argument may vary from a cycle to another in sequential data processing, but NOBL, M, and L should be kept constant. For efficiency, it is advisable to use NSMP as large as possible.
- When 100 cycles of sequential data processing are completed with LAST = .FALSE., a warning is issued, to prevent for an infinite loop.

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Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

Consider the discrete-time system (A,B,C,D) with the following matrices:

$$A = \begin{pmatrix} 1.5 & -0.7 \\ 1.0 & 0.0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0.5 \end{pmatrix} \quad D = 0$$

whose output response $y(t)$ to random input signals $u(t)$ and zero initial state is determined by using the subroutine RPIMIU. The following sequence of statements can be used to estimate the order and the matrices of the original system

```

DATA   ISEED/ 1, 2, 3, 5 /
NSMP   = 120
NOBL   = 8
M      = 1
L      = 1
TOL    = -1.0D0
WITHD  = .FALSE.
LAST   = .TRUE.
CONCT  = .TRUE.
CTRL   = .TRUE.
LRWORK = 8100
C      Compute the output response for uniformly distributed
C      input sequence.

```

```

X(1) = 0.000
X(2) = 0.000
CALL RPIMIUI( NSMP, 2, M, L, 0, 0, A, B, E, C, D, F, X, U,
*           W, V, ISEED, Y, RWORK, LRWORK, WITHD, 'G',
*           'G', 'R', 'R', 'R', 1, *1111 )
C   Perform the system identification.
CALL RPIMPI( U, NSMP, M, Y, L, NOBL, TOL, WITHD, CONCT, LAST,
*           CTRL, N, A, B, C, D, R, IWORK, RWORK, LRWORK,
*           *1111 )

```

The order detection singular values are:

```
2.7731D+01  1.2852D+01  0  0  0  0  0  0
```

The estimated system matrices are:

$$\bar{A} = \begin{pmatrix} .7923 & -.5521 \\ .2523 & .7077 \end{pmatrix} \quad \bar{B} = \begin{pmatrix} 3.4759 \\ -2.2976 \end{pmatrix} \quad \bar{C} = \begin{pmatrix} .5794 & .4413 \end{pmatrix}$$

which coincide with the matrices obtained by applying a similarity transformation to the original system.

The relative output error is: $\|y(t) - \bar{y}(t)\| / \|y(t)\| = 0.80622D-15$

Error Messages:

```

-1-
Invalid parameter value on entry.
-2-
Singular value decomposition failed.
-3-
A singular upper triangular matrix was found.
-4-
Not enough working storage. It should be at least //LENG//.

```

Warnings Messages:

```

-1-
More than 100 cycles in sequential data processing
-2-
All singular values are exactly zero (N = 0)

```

SUBROUTINE RPIMPO

Ordinary MOESP identification with past inputs and outputs

Procedure purpose:

This subroutine estimates from given input and output data sequences, the order and the matrices of the linear time-invariant discrete-time state space model

$$x(k+1) = Ax(k) + Bu(k) + w(k)$$

$$y(k) = Cx(k) + Du(k) + v(k)$$

by using the ordinary MOESP algorithm with past inputs and outputs.

Usage:

```
CALL RPIMPO(U, NSMP, M, Y, L, NOBL, TOL, WITHD, CONCT, LAST,
           CTRL, N, A, B, C, D, R, IWORK, RWORK, LRWORK, *)
```

U : IN, DOUBLE (NSMP,M)
 system input data sequence matrix $U = [u_1 \ u_2 \ \dots \ u_m]$.
 Column j of U contains the NSMP values of the j -th input component for consecutive time increments.

NSMP : IN, INTEGER
 number of rows of matrices U and Y (number of samples).
 (When sequential data processing is used, NSMP is the number of samples of the current data batch.)
 NSMP $\geq 2*(M+L+1)*NOBL - 1$, for non-sequential processing of the data;
 NSMP $\geq 2*NOBL$, for sequential processing of several data batches;
 NSMP = 0, and LAST = .TRUE., is allowed if the upper triangular factor of the QR factorization is already available in the array R , and only the system matrices are to be computed.
 The total number of samples when calling with LAST = .TRUE. should be at least $2*(M+L+1)*NOBL - 1$.

M : IN, INTEGER
 dimension of system input vector.

Y : IN, DOUBLE (NSMP,L)
 system output data sequence matrix $Y = [y_1 \ y_2 \ \dots \ y_l]$.
 Column j of Y contains the NSMP values of the j -th output component for consecutive time increments.

L : IN, INTEGER
 dimension of system output vector.

NOBL : IN, INTEGER
number of block rows in the processed input and output
Hankel matrices. NOBL should be chosen larger than n ,
the estimated dimension of state vector.

TOL : IN, DOUBLE
absolute tolerance used for determining an estimate of the
system order. If $TOL > 0$, the estimate is indicated by the
number of singular values greater than or equal to TOL .
(Singular values less than TOL are considered zero.)
If $TOL = 0$, the internally computed default value
 $TOL = NOBL * EPS * SV[1]$ is used, where EPS is the relative
machine precision, and $SV[1]$ is the maximal singular value.
If $TOL < 0$, the estimate is indicated by the index
of the singular value that has the largest logarithmic
gap to its successor. (See Method for the definition of
the singular values.)

WITHD : IN, LOGICAL
specifies whether or not a non-zero feedthrough matrix D
to be included in the estimated state space representation:
WITHD = .TRUE. means a possibly non-zero D matrix should be
estimated.
WITHD = .FALSE. means a zero D matrix is assumed.

CONCT : IN, LOGICAL
specifies whether or not the successive data batches belong
to a single experiment:
CONCT = .TRUE. means the current data batch is a
continuation of the previous data batch.
CONCT = .FALSE. means there is no connection between the
current data batch and the previous ones.

LAST : IN, LOGICAL
specifies whether or not sequential data processing is
used, and (for sequential processing) whether or not the
current data block is the last one, as follows:
LAST = .TRUE. means non-sequential processing or the last
data block in sequential processing.
LAST = .FALSE. means not the last data block in sequential
processing. In this case the only calculation
performed is to update the triangular factor
of the QR factorization.

CTRL : IN, LOGICAL
specifies whether or not the user's confirmation of the
system order estimate is desired before the computation
of system matrices, as follows:
CTRL = .TRUE. means user's confirmation is desired.
CTRL = .FALSE. means no confirmation is necessary.
If CTRL = .TRUE., a reverse communication routine,
IMCORD, is called, and, after inspecting the singular

values and system order estimate, n , the user may accept n or set a new value.
 IMCORD is not called by the routine if CTRL = .FALSE..
 CTRL = .TRUE. should be used in extreme cases only.

N : OUT, INTEGER
 estimated order of the system.

A : OUT, DOUBLE (NOBL-1,NOBL-1)
 the $N \times N$ estimated system state matrix A.
 (column dense)

B : OUT, DOUBLE (NOBL-1,M)
 the $N \times M$ estimated system input matrix B.
 (column dense)

C : OUT, DOUBLE (L,NOBL-1)
 the $L \times N$ estimated system output matrix C.
 (column dense)

D : OUT, DOUBLE (L,M)
 the $L \times M$ estimated system feedthrough matrix D if
 WITHD = .TRUE..
 D is not referenced if WITHD = .FALSE..
 (column dense)

R : IN, OUT, DOUBLE (max(2*(M+L)*NOBL, 3*M*NOBL),2*(M+L)*NOBL)
 On output, if LAST = .FALSE., then the leading
 $2*(M+L)*NOBL \times 2*(M+L)*NOBL$ part of R contains the current
 upper triangular factor of the QR factorization used to
 compress the data before determining the singular values
 in the MOESP algorithm.
 The content of R should be preserved between successive
 calls of RPIMPO with LAST = .FALSE..
 On first input, the content of R is not meaningful.

IWORK : OUT, INTEGER(M)
 working array.

RWORK : OUT, DOUBLE (LRWORK)
 working array.
 RWORK(1) contains the optimal suggested value of LRWORK;
 RWORK(2) contains an estimate of the reciprocal condition
 number for the final linear algebraic system solved;
 RWORK(3:2+L*NOBL) contain the singular values used
 for determining the system order.
 The first $(M+L)*(2*NOBL-1)$ elements of RWORK should be
 preserved between successive calls of the routine with
 LAST = .FALSE., and the final call with LAST = .TRUE.,
 if successive data batches belong to a single experiment
 (CONCT = .TRUE.).
 On return with error number 4, RWORK(1) contains the
 minimum necessary value of LRWORK.

LRWORK: IN, INTEGER
 dimension of working array RWORK.

The value of LRWORK must be at least LRWMIN, where

$$\text{LRWMIN} = \max(4*(\text{NOBL}+1)*(M+L)*\text{NOBL}, 2+L*\text{NOBL} +$$

$$\max(((2*\text{NOBL}-3)*L+2)*(\text{NOBL}-1),$$

$$(L*\text{NOBL}-1)*M*\text{NOBL}, 4*(L*\text{NOBL}+M),$$

$$5*(L*\text{NOBL}+M)-4))$$

For NSMP = 0, the first term of the first max expression could be omitted.

This estimate is computed using the largest and smallest values for n, namely NOBL-1 and 1, respectively (for positive and negative terms, respectively).

In practice, n could be much smaller than NOBL-1.

For good performance, LRWORK should be larger.

* : RETURN 1, target label in case of error (e.g. *1111)

File input/ output:

none

Method:

Let us denote $m = M$, $l = L$, $t = \text{NSPM}$, $s = \text{NOBL}$.

For non-sequential data processing, the $t \times 2(m+1)s$ matrix

$$\begin{bmatrix} \text{Uf}' & \text{Up}' & \text{Y}' \\ s+1,s,t & 1,s,t & 1,2s,t \end{bmatrix}$$

is constructed, where Up , Uf and Y are $1,s,t$, $s+1,s,t$ and $1,2s,t$

Hankel matrices defined in terms of the input and output data /3/.

A QR factorization is used to compress the data, and then a singular value decomposition (SVD) of the submatrix

$[\text{R}_{24}' \ \text{R}_{34}']' := \text{R}(ms+1:(2m+1)s, (2m+1)s+1:2(m+1)s)$ of the upper triangular factor R reveals the order n of the system as the number of "non-zero" singular values. System matrices are finally computed from the right singular vectors of $[\text{R}_{24}' \ \text{R}_{34}']'$ and the final, "compressed" submatrices R_{11} and R_{12} of R .

For sequential data processing, the QR decomposition is done sequentially, by updating the upper triangular factor R . When all data have been compressed, the system order and system matrices are computed as in the previous case.

Literature

/1/ Verhaegen M., and P. Dewilde

Subspace Model Identification. Part 1: The output-error state-space model identification class of algorithms.

Int. J. Control, 56, pp. 1187-1210, 1992.

/2/ Verhaegen M.

Subspace Model Identification. Part 3: Analysis of the ordinary output-error state-space model identification algorithm.

Int. J. Control, 58, pp. 555-586, 1993.

/3/ Verhaegen M.

Identification of the deterministic part of MIMO state space models given in innovations form from input-output data.

Automatica, 30, pp. 61-74, 1994.

Remarks:

- The NSMP argument may vary from a cycle to another in sequential data processing, but NOBL, M, and L should be kept constant. For efficiency, it is advisable to use NSMP as large as possible.
- When 100 cycles of sequential data processing are completed with LAST = .FALSE., a warning is issued, to prevent for an infinite loop.

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Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

Consider the discrete-time system (A,B,C,D) with the following matrices:

$$A = \begin{pmatrix} 1.5 & -0.7 \\ 1.0 & 0.0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 2.5 & -0.2 \end{pmatrix} \quad D = \begin{pmatrix} 1 \end{pmatrix}$$

whose output response $y(t)$ to random input signals $u(t)$ and zero initial state is determined by using the subroutine RPIMIU.

The following sequence of statements can be used to estimate the order and the matrices of the original system

```
DATA  ISEED/ 1, 2, 3, 5 /
NSMP  = 80
NOBL  = 6
M     = 1
L     = 1
TOL   = -1.000
WITHD = .TRUE.
CONCT = .TRUE.
LAST  = .TRUE.
```

```

CTRL = .TRUE.
LRWORK = 8100
C   Compute the output response for uniformly distributed
C   input sequence.
X(1) = 0.0D0
X(2) = 0.0D0
CALL RPIMIUI( NSMP, 2, M, L, 0, 0, A, B, E, C, D, F, X, U,
*           W, V, ISEED, Y, RWORK, LRWORK, WITHD, 'G',
*           'G', 'R', 'R', 'R', 1, *1111 )
C   Perform the system identification.
CALL RPIMPO( U, NSMP, M, Y, L, NOBL, TOL, WITHD, CONCT, LAST,
*           CTRL, N, A, B, C, D, R, IWORK, RWORK, LRWORK,
*           *1111 )

```

The order detection singular values are:

```
3.5290D+01  1.5322D+01  0  0  0  0
```

The estimated system matrices are:

$$\bar{A} = \begin{pmatrix} .7236 & .5580 \\ -.2476 & .7764 \end{pmatrix} \quad \bar{B} = \begin{pmatrix} 5.6737 \\ 3.3928 \end{pmatrix} \quad \bar{C} = \begin{pmatrix} .6706 & -.3845 \end{pmatrix}$$

$$\bar{D} = \begin{pmatrix} 1 \end{pmatrix}$$

which coincide with the matrices obtained by applying a similarity transformation to the original system.

The relative output error is: $\|y(t) - \bar{y}(t)\| / \|y(t)\| = 0.89731D-15$

Error Messages:

```

-1-
Invalid parameter value on entry.
-2-
Singular value decomposition failed.
-3-
A singular upper triangular matrix was found.
-4-
Not enough working storage. It should be at least //LENG//.

```

Warnings Messages:

```

-1-
More than 100 cycles in sequential data processing
-2-
All singular values are exactly zero (N = 0)

```

SUBROUTINE RPIMRS

Ordinary MOESP Identification with Reconstructed States and Past Inputs

Procedure purpose:

This subroutine determines the order of a linear time-invariant discrete-time state space model and then estimates the quadruple of system matrices using the MOESP algorithm with reconstructed state variables as instrumental variables.

The model structure is :

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) \\y(k) &= Cx(k) + Du(k) + e(k)\end{aligned}$$

where $e(k)$ is a zero-mean noise of arbitrary color, independent of the noise-free input $u(k)$. Matrix D could be zero.

Usage:

```
CALL RPIMRS( U, NSMP, M, Y, L, XEST, N, NOBL, TOL, WITHD, CONCT,
            LAST, CTRL, A, B, C, D, R, IWORK, RWORK, LRWORK, * )
```

U : IN, DOUBLE (NSMP,M)
system input data sequence matrix $U = [u_1 \ u_2 \ \dots \ u_m]$.
Column j of U contains the NSMP values of the
 j -th input component for consecutive time increments.

NSMP : IN, INTEGER
number of rows of matrices U , Y and $XEST$ (number of samples)
(When sequential data processing is used, NSMP is the
number of samples of the current data batch.)
NSMP $\geq (M+L+1)*NOBL + N - 1$, for non-sequential processing
of the data;
NSMP $\geq NOBL$, for sequential processing of several data batches;
NSMP = 0, and LAST = .TRUE., is allowed if the upper
triangular factor of the QR factorization is
already available in the array R , and only the
system matrices are to be computed.
The total number of samples when calling with LAST = .TRUE.
should be at least $(M+L+1)*NOBL + N - 1$.

M : IN, INTEGER
dimension of system input vector.

Y : IN, DOUBLE (NSMP,L)
system output data sequence matrix $Y = [y_1 \ y_2 \ \dots \ y_l]$.
Column j of Y contains the NSMP values of the
 j -th output component for consecutive time increments.

L : IN, INTEGER
 dimension of system output vector.

XEST : IN, DOUBLE (NSMP,N)
 estimated state sequence matrix $XEST = [x_1 \ x_2 \ \dots \ x_n]$.
 Column j of XEST contains the NSMP values of the j -th
 estimated state component for consecutive time increments.

N : IN, OUT, INTEGER
 estimated order of the system.
 On input, it is possible to use $N = \max(1, NOBL-1)$.

NOBL : IN, INTEGER
 number of block rows in the processed input and output
 Hankel matrices. NOBL should be chosen larger than n ,
 the estimated dimension of state vector.

TOL : IN, DOUBLE
 absolute tolerance used for determining an estimate of the
 system order. If $TOL > 0$, the estimate is indicated by the
 number of singular values greater than or equal to TOL.
 (Singular values less than TOL are considered zero.)
 If $TOL = 0$, the internally computed default value
 $TOL = NOBL * EPS * SV[1]$ is used, where EPS is the relative
 machine precision, and SV[1] is the maximal singular value.
 If $TOL < 0$, the estimate is indicated by the index
 of the singular value that has the largest logarithmic
 gap to its successor. (See Method for the definition of
 the singular values.)

WITHD : IN, LOGICAL
 specifies whether or not a non-zero feedthrough matrix D
 to be included in the estimated state space representation:
 WITHD = .TRUE. means a possibly non-zero D matrix should be
 estimated.
 WITHD = .FALSE. means a zero D matrix is assumed.

CONCT : IN, LOGICAL
 specifies whether or not the successive data batches belong
 to a single experiment:
 CONCT = .TRUE. means the current data batch is a
 continuation of the previous data batch.
 CONCT = .FALSE. means there is no connection between the
 current data batch and the previous ones.

LAST : IN, LOGICAL
 specifies whether or not sequential data processing is
 used, and (for sequential processing) whether or not the
 current data block is the last one, as follows:
 LAST = .TRUE. means non-sequential processing or the last
 data block in sequential processing.
 LAST = .FALSE. means not the last data block in sequential
 processing. In this case the only calculation
 performed is to update the triangular factor

of the QR factorization.

CTRL : IN, LOGICAL
specifies whether or not the user's confirmation of the system order estimate is desired before the computation of system matrices, as follows:
CTRL = .TRUE. means user's confirmation is desired.
CTRL = .FALSE. means no confirmation is necessary.
If CTRL = .TRUE., a reverse communication routine, IMCORD, is called, and, after inspecting the singular values and system order estimate, n , the user may accept n or set a new value.
IMCORD is not called by the routine if CTRL = .FALSE..
CTRL = .TRUE. should be used in extreme cases only.

A : OUT, DOUBLE (NOBL-1,NOBL-1)
the $N \times N$ estimated system state matrix A.
(column dense)

B : OUT, DOUBLE (NOBL-1,M)
the $N \times M$ estimated system input matrix B.
(column dense)

C : OUT, DOUBLE (L,NOBL-1)
the $L \times N$ estimated system output matrix C.
(column dense)

D : OUT, DOUBLE (L,M)
the $L \times M$ estimated system feedthrough matrix D if WITHD = .TRUE.. D is not referenced if WITHD = .FALSE..
(column dense)

R : IN, OUT, DOUBLE (max(M*NOBL+N,L*NOBL)+L*NOBL,(M+L)*NOBL+N)
On output, if LAST = .FALSE., then the leading $(M+L) \times NOBL + N$ x $(M+L) \times NOBL + N$ part of R contains the current upper triangular factor of the QR factorization used to compress the data before determining the singular values in the MOESP algorithm.
The content of R should be preserved between successive calls of RPIMRS with LAST = .FALSE..
On first input, the content of R is not meaningful.

IWORK : OUT, INTEGER(M)
working array.

RWORK : OUT, DOUBLE (LRWORK)
working array.
RWORK(1) contains the optimal suggested value of LRWORK;
RWORK(2) contains an estimate of the reciprocal condition number for the final linear algebraic system solved;
RWORK(3:2+L*NOBL) contain the singular values used for determining the system order.
The first $(M+N+L) \times (NOBL-1)$ elements of RWORK should be preserved between successive calls of the routine with LAST = .FALSE., and the final call with LAST = .TRUE.,

if successive data batches belong to a single experiment
(CONCT = .TRUE.).

On return with error number 4, RWORK(1) contains the
minimum necessary value of LRWORK.

LRWORK: IN, INTEGER

dimension of working array RWORK.

The value of LRWORK must be at least LRWMIN, where

$$\text{LRWMIN} = \max((\text{NOBL}+2)*((\text{M}+\text{L})*\text{NOBL}+\text{N}), \\ 2+\text{L}*\text{NOBL} + \max(((2*\text{NOBL}-3)*\text{L}+2)*\text{N}, \\ (\text{L}*\text{NOBL}-\text{N})*\text{M}*\text{NOBL}, \\ 4*(\text{L}*\text{NOBL}+\text{M}), 5*(\text{L}*\text{NOBL}+\text{M})-4))$$

For NSMP = 0, the first term of the first max expression
could be omitted.

This estimate is computed using the given value for n.

For good performance, LRWORK should be larger.

* : RETURN 1, target label in case of error (e.g. *1111)

File input/ output:

none

Method:

Let us denote $m = M$, $l = L$, $t = \text{NSPM}$, $s = \text{NOBL}$.

For non-sequential data processing, the $t \times (m+1)s+n$ matrix

$$\begin{bmatrix} U' & Xest' & Y' \\ 1,s,t & & 1,s,t \end{bmatrix}$$

is constructed, where U and Y are Hankel matrices
 $1,s,t$ $1,s,t$

defined in terms of the input and output data /1/, /2/ and Xest
is the estimated state trajectory.

A QR factorization is used to compress the data, and then a
singular value decomposition (SVD) of the submatrix

R_23 := R(ms+1:ms+n,ms+n+1:(m+1)s+n) of the upper triangular factor
R reveals the order n of the system as the number of "non-zero"
singular values. System matrices are finally computed from the right
singular vectors of R_23 and the submatrices R_11 and R_13 of R.

For sequential data processing, the QR decomposition is done
sequentially, by updating the upper triangular factor R. When
all data have been compressed, the system order and system
matrices are computed as in the previous case.

Literature

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/2/ Verhaegen M.

Subspace Model Identification. Part 3: Analysis of the ordinary output-error state-space model identification algorithm.

Int. J. Control, 58, pp. 555-586, 1993.

Remarks:

- The NSMP argument may vary from a cycle to another in sequential data processing, but NOBL, N, M, and L should be kept constant. For efficiency, it is advisable to use NSMP as large as possible.
- When 100 cycles of sequential data processing are completed with LAST = .FALSE., a warning is issued, to prevent for an infinite loop.

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Life cycle:

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Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

Consider the discrete-time system (A,B,C,D) with the following matrices:

$$A = \begin{pmatrix} 1.5 & -0.7 \\ 1.0 & 0.0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0.5 \end{pmatrix} \quad D = 0$$

whose output response $y(t)$ to random input signals $u(t)$ and zero initial state is determined by using the subroutine RPIMIU.

The following sequence of statements can be used to estimate the order and the matrices of the original system

```

DATA   ISEED/ 1, 2, 3, 5 /
NSMP   = 120
NOBL   = 10
N      = 2
M      = 1
L      = 1
TOL    = -1.0D0
WITHD  = .FALSE.
LAST   = .TRUE.
CONCT  = .TRUE.
CTRL   = .TRUE.
LRWORK = 8100
C      Compute the output response for uniformly distributed

```



```

C   input sequence.
      X(1) = 0.0D0
      X(2) = 0.0D0
      CALL RPIMIU( NSMP, N, M, L, 0, 0, A, B, E, C, D, F, X, U,
*                W, V, ISEED, Y, RWORK, LRWORK, WITHD, 'G',
*                'G', 'R', 'R', 'R', 1, *1111 )
C   Compute system state trajectory.
      CI = 2 x 2 identity matrix
      CALL RPIMIU( NSMP, N, M, N, 0, 0, A, B, E, CI, D, F, X, U,
*                W, V, ISEED, XEST, RWORK, LRWORK, .FALSE., 'G',
*                'G', 'G', 'R', 'R', 1, *1111 )
C   Perform the system identification.
      CALL RPIMRS( U, NSMP, M, Y, L, XEST, N, NOBL, TOL, WITHD,
*                CONCT, LAST, CTRL, A, B, C, D, R, IWORK,
*                RWORK, LRWORK, *1111 )

```

The order detection singular values are:

```
3.2203D+01  1.4612D+01  0  0  0  0  0  0  0  0
```

The estimated system matrices are:

$$\bar{A} = \begin{pmatrix} .7831 & .5606 \\ -.2472 & .7169 \end{pmatrix} \quad \bar{B} = \begin{pmatrix} 3.5468 \\ 2.3707 \end{pmatrix} \quad \bar{C} = \begin{pmatrix} .5749 & -.4382 \end{pmatrix}$$

which coincide with the matrices obtained by applying a similarity transformation to the original system.

The relative output error is: $\|y(t) - \bar{y}(t)\| / \|y(t)\| = 0.95410D-15$

Error Messages:

-1-

Invalid parameter value on entry.

-2-

Singular value decomposition failed.

-3-

A singular upper triangular matrix was found.

-4-

Not enough working storage. It should be at least //LENG//.

Warnings Messages:

-1-

More than 100 cycles in sequential data processing

-2-

All singular values are exactly zero (N = 0)

SUBROUTINE RPIMKG

MOESP-SI Deterministic-Stochastic Identification

Procedure purpose:

This subroutine determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices and, optionally, the noise covariance matrices and the Kalman gain matrix using the "ordinary MOESP algorithm with past inputs and outputs", combined with the "subspace state space system identification" algorithm, for covariance calculations. It is assumed that the input is a zero-mean white noise process.

The model structure is :

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) + w(k) \\y(k) &= Cx(k) + Du(k) + e(k)\end{aligned}$$

where $w(k)$ and $e(k)$ are zero-mean white noise sequences, independent of the input $u(k)$. Matrix D could be zero.

The Kalman predictor structure is :

$$\begin{aligned}\bar{x}(k+1) &= A\bar{x}(k) + Bu(k) + K(y(k) - C\bar{x}(k) - Du(k)) \\ \bar{y}(k) &= C\bar{x}(k) + Du(k)\end{aligned}$$

where K is the Kalman gain matrix corresponding to the estimated noise covariance matrices.

There is an option for sequential input-output data processing.

Usage:

```
CALL RPIMKG( U, NSMP, M, Y, L, NOBL, TOL, WITHD, WITHCV, WITHK, CONCT,
            LAST, CTRL, N, A, B, C, D, FK, R, IWORK, RWORK, LRWORK, *)
```

U : IN, DOUBLE (NSMP,M)
system input data sequence matrix $U = [u_1 \ u_2 \ \dots \ u_m]$.
Column j of U contains the NSMP values of the j -th input component for consecutive time increments.

NSMP : IN, INTEGER
number of rows of matrices U and Y (number of samples)
(When sequential data processing is used, NSMP is the number of samples of the current data batch.)
NSMP $\geq 2*(M+L+1)*NOBL - 1$, for non-sequential processing of the data;

NSMP $\geq 2 \cdot \text{NOBL}$, for sequential processing of several data batches;

NSMP = 0, and LAST = .TRUE., is allowed if the upper triangular factor of the QR factorization is already available in the array R, and only the system matrices are to be computed.

The total number of samples when calling with LAST = .TRUE. should be at least $2 \cdot (M+L+1) \cdot \text{NOBL} - 1$.

- M : IN, INTEGER
dimension of system input vector.
- Y : IN, DOUBLE (NSMP,L)
system output data sequence matrix $Y = [y_1 \ y_2 \ \dots \ y_l]$.
Column j of Y contains the NSMP values of the j-th output component for consecutive time increments.
- L : IN, INTEGER
dimension of system output vector.
- NOBL : IN, INTEGER
number of block rows in the processed input and output Hankel matrices. NOBL should be chosen larger than n, the estimated dimension of state vector.
- TOL : IN, DOUBLE
absolute tolerance used for determining an estimate of the system order. If TOL > 0, the estimate is indicated by the number of singular values greater than or equal to TOL. (Singular values less than TOL are considered zero.)
If TOL = 0, the internally computed default value $\text{TOL} = \text{NOBL} \cdot \text{EPS} \cdot \text{SV}[1]$ is used, where EPS is the relative machine precision, and SV[1] is the maximal singular value.
If TOL < 0, the estimate is indicated by the index of the singular value that has the largest logarithmic gap to its successor. (See Method for the definition of the singular values.)
- WITHD : IN, LOGICAL
specifies whether or not a non-zero feedthrough matrix D to be included in the estimated state space representation:
WITHD = .TRUE. means a possibly non-zero D matrix should be estimated.
WITHD = .FALSE. means a zero D matrix is assumed.
- WITHCV: IN, LOGICAL
specifies whether or not the covariance matrices are to be computed:
WITHCV = .TRUE. means the covariance matrices are computed.
WITHCV = .FALSE. means the covariance matrices should not be computed.
The covariance matrices are automatically computed if WITHK = .TRUE..

WITHK : IN, LOGICAL
 specifies whether or not the Kalman gain matrix K is to be computed:
WITHK = .TRUE. means the Kalman gain matrix is computed.
WITHK = .FALSE. means the Kalman gain matrix is not computed.

CONCT : IN, LOGICAL
 specifies whether or not the successive data batches belong to a single experiment:
CONCT = .TRUE. means the current data batch is a continuation of the previous data batch.
CONCT = .FALSE. means there is no connection between the current data batch and the previous ones.

LAST : IN, LOGICAL
 specifies whether or not sequential data processing is used, and (for sequential processing) whether or not the current data block is the last one, as follows:
LAST = .TRUE. means non-sequential processing or the last data block in sequential processing.
LAST = .FALSE. means not the last data block in sequential processing. In this case the only calculation performed is to update the triangular factor of the QR factorization.

CTRL : IN, LOGICAL
 specifies whether or not the user's confirmation of the system order estimate is desired before the computation of system matrices, as follows:
CTRL = .TRUE. means user's confirmation is desired.
CTRL = .FALSE. means no confirmation is necessary.
 If **CTRL** = .TRUE., a reverse communication routine, IMCORD, is called, and, after inspecting the singular values and system order estimate, n , the user may accept n or set a new value.
 IMCORD is not called by the routine if **CTRL** = .FALSE..
CTRL = .TRUE. should be used in extreme cases only.

N : OUT, INTEGER
 estimated order of the system.

A : OUT, DOUBLE (NOBL-1,NOBL-1)
 the $N \times N$ estimated system state matrix A .
 (column dense)

B : OUT, DOUBLE (NOBL-1,M)
 the $N \times M$ estimated system input matrix B .
 (column dense)

C : OUT, DOUBLE (L,NOBL-1)
 the $L \times N$ estimated system output matrix C .
 (column dense)

D : OUT, DOUBLE (L,M)

the $L \times M$ estimated system feedthrough matrix D if
 WITHD = .TRUE..
 D is not referenced if WITHD = .FALSE..
 (column dense)

FK : OUT, DOUBLE (NOBL-1,L)
 if WITHK = .TRUE., on normal return, or when the
 warning errors 1 or 3 are issued, the leading $N \times L$ part
 of this array contains the Kalman gain matrix K .
 FK is not referenced when WITHK = .FALSE..
 (column dense)

R : IN, OUT, DOUBLE (max(2*(M+L)*NOBL, 3*M*NOBL),2*(M+L)*NOBL)
 On output, if LAST = .FALSE., then the leading
 $2*(M+L)*NOBL \times 2*(M+L)*NOBL$ part of R contains the
 current upper triangular factor of the QR factorization
 used to compress the data before determining the singular
 values in the MOESP algorithm.
 The content of R should be preserved between successive
 calls of RPIMKG with LAST = .FALSE..
 On first input, the content of R is not meaningful.
 During the computations with LAST = .TRUE., the leading
 $M*NOBL \times (M+L)*NOBL$ part of R contains some relevant
 data for the MOESP algorithm (the concatenated
 compressed matrices $R_{11} = R_{1C}'$, and $R_{12} = R_{2C}'$).
 On output with LAST = .TRUE., the content of R is
 destroyed; if WITHCV = .TRUE. or WITHK = .TRUE.,
 on normal return, and in absence of the warning error 2,
 the first $N+L$ rows and columns of R contain the estimated
 covariance matrix of the vector $[w', e']'$;
 if WITHK = .TRUE., on normal return, or when the warning
 errors 1 or 3 are issued, the rows $N+L+1$ to $2*N+L+1$
 and the columns 1 to N of R contain the estimation error
 covariance matrix X (the Riccati equation solution).

IWORK : OUT, INTEGER(K)
 working array.
 $K = M$, if WITHCV = .FALSE. and WITHK = .FALSE.
 $K = M*NOBL + NOBL - 1$, if WITHCV = .TRUE., or WITHK = .TRUE..

RWORK : OUT, DOUBLE (LRWORK)
 working array.
 RWORK(1) contains the optimal suggested value of LRWORK;
 RWORK(2) contains an estimate of the reciprocal condition
 number for the final linear algebraic system solved for
 computing system matrices (in the absence of warning error 2);
 RWORK(3:L*NOBL+2) contain singular values used for
 determining the system order;
 If WITHK = .TRUE., LAST = .TRUE., on normal return,
 or when the warning errors 1 or 3 are issued, then:
 RWORK(L*NOBL+3) contains an estimate of the reciprocal

condition number of the computed output covariance matrix R_y ;
 RWORK(L*NOBL+4) contains an estimate of the reciprocal
 condition number of the computed matrix A ;
 RWORK(L*NOBL+5) contains an estimate of the reciprocal
 condition number of the computed matrix U_{11} , used to
 obtain the Riccati equation solution X ;
 RWORK(L*NOBL+6) contains an estimate of the reciprocal
 condition number of the matrix $R_y + C*X*C'$, used to
 obtain the Kalman gain matrix;
 RWORK(L*NOBL+7:L*NOBL+N+6) contain the real parts of the
 optimal (Kalman predictor) poles;
 RWORK(L*NOBL+N+7:L*NOBL+2*N+6) contain the imaginary
 parts of the optimal (Kalman predictor) poles.
 Part of this information is also available for warning
 errors 4 to 9, if logically possible.
 The first $(M+L)*(2*NOBL - 1)$ elements of RWORK should
 be preserved between successive calls of the routine with
 LAST = .FALSE., and the final call with LAST = .TRUE.,
 if successive data batches belong to a single experiment
 (CONCT = .TRUE.).

On return with error number 4, RWORK(1) contains the
 minimum necessary value of LRWORK.

LRWORK: IN, INTEGER

dimension of working array RWORK.

The value of LRWORK must be at least LRWMIN, where

$$LRWMIN = \max(4*(NOBL+1)*(M+L)*NOBL, 2+L*NOBL + \\
\max(((2*NOBL-3)*L+2)*(NOBL-1), \\
(L*NOBL-1)*M*NOBL, 4*(L*NOBL+M), \\
5*(L*NOBL+M)-4, \\
3 + \max(10*(NOBL-1), 3*L), \\
5 + 2*(NOBL-1) + \max(NOBL-1, 3*L)))$$

For NSMP = 0, the first term of the first max expression
 could be omitted. For WITHK = .FALSE., the last two terms
 of the second max expression could be omitted.

This estimate is computed using the largest and smallest
 values for n , namely $NOBL-1$ and 1 , respectively (for
 positive and negative terms, respectively).

For good performance, LRWORK should be larger.

* : RETURN 1, target label in case of error (e.g. *1111)

File input/ output:

none

Method:

Let us denote $m = M$, $l = L$, $t = NSPM$, $s = NOBL$.

1) For non-sequential data processing, the $t * 2(m+1)s$ matrix

$$\begin{bmatrix} U_f' & U_p' & Y' \\ s+1,s,t & 1,s,t & 1,2s,t \end{bmatrix}$$

is constructed, where U_p , U_f and Y are Hankel matrices defined in terms of the input and output data [3]. A QR factorization is used to compress the data, and then a singular value decomposition (SVD) of the submatrix $[R_{24}' \ R_{34}']' := R(ms+1:(2m+1)s, (2m+1)s+1:2(m+1)s)$ of the upper triangular factor R reveals the order n of the system as the number of "non-zero" singular values. System matrices are finally computed using the right singular vectors of $[R_{24}' \ R_{34}']'$ and the final, "compressed" submatrices R_{11} and R_{12} of R .

2) For sequential data processing, the QR decomposition is done sequentially, by updating the upper triangular factor R . When all data have been compressed, the system order and system matrices are computed as in the previous case.

The covariance matrices are computed by solving several least squares problems, constructed based on the N4SID approach [4].

The Kalman gain is obtained by solving a discrete-time algebraic matrix Riccati equation using the Schur vectors approach [5] for the dual of an optimal control problem.

Literature

- /1/ Verhaegen M., and P. Dewilde
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- /3/ Verhaegen M.
Identification of the deterministic part of MIMO state space models given in innovations form from input-output data.
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- /4/ Van Overschee, P., and De Moor, B.
N4SID: Two Subspace Algorithms for the Identification of Combined Deterministic-Stochastic Systems.
Automatica, Vol.30, No.1, pp. 75-93, 1994.

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A Schur method for solving algebraic Riccati equations.
IEEE Trans. Automat. Control, AC-24, pp. 913-921, 1979.

Remarks:

- The NSMP argument may vary from a cycle to another in sequential data processing, but NOBL, M, and L should be kept constant. For efficiency, it is advisable to use NSMP as large as possible.
- When 100 cycles of sequential data processing are completed with LAST = .FALSE., a warning is issued, to prevent for an infinite loop.

Copyright:

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Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

Consider the discrete-time system

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) + Ew(k) \\y(k) &= Cx(k) + Du(k) + Fv(k)\end{aligned}$$

with the following matrices:

$$A = \begin{pmatrix} 1.5 & -0.7 \\ 1.0 & 0.0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0.5 \end{pmatrix} \quad D = 0$$

$$E = \begin{pmatrix} 0.0005 & 0.0 \\ 0.0 & 0.006 \end{pmatrix} \quad F = \begin{pmatrix} 0.0528 \end{pmatrix}$$

whose output response $y(t)$ to normal $(0,1)$ random input signals $u(t)$, $w(t)$, $v(t)$ and zero initial state is determined by using the subroutine RPIMIU.

The following sequence of statements can be used to estimate the order and the matrices of the original system

```
DATA ISEED/ 1, 2, 3, 5 /
NSMP = 120
NOBL = 6
M = 1
MW = 2
L = 1
```



```

LV      = 1
TOL     = -1.0D0
WITHD   = .FALSE.
WITHK   = .TRUE.
LAST    = .TRUE.
CONCT   = .TRUE.
CTRL    = .TRUE.
LRWORK  = 8100
C       Compute the output response for uniformly distributed
C       input sequence.
X(1)    = 0.0D0
X(2)    = 0.0D0
CALL RPIMIU( NSMP, 2, M, L, MW, LV, A, B, E, C, D, F, X, U,
*           W, V, ISEED, Y, RWORK, LRWORK, WITHD, 'G',
*           'G', 'R', 'R', 'R', 3, *1111 )
C       Perform the system identification.
CALL RPIMKG( U, NSMP, M, Y, L, NOBL, TOL, WITHD, WITHCV,
*           WITHK, CONCT, LAST, CTRL, N, A, B, C, D, FK,
*           R, IWORK, RWORK, LRWORK, *1111 )

```

The order detection singular values are:

```

.87524260D+02  .42090540D+02  .32603442D+00  .22092251D+00  .19614963D+00
.16624421D+00

```

We can take the system order $n = 2$.

The estimated system matrices are:

```

 $\bar{A} = \begin{pmatrix} .8245 & .5418 \\ -.2643 & .6748 \end{pmatrix}$ 
 $\bar{B} = \begin{pmatrix} 3.8186 \\ 1.6312 \end{pmatrix}$ 
 $\bar{C} = \begin{pmatrix} .5093 & -.5822 \end{pmatrix}$ 

```

The estimated Kalman gain matrix K is:

```

K = ( .1190 )
     (-.1635 )

```

The estimated (inter-)covariance matrix (for $[Ew; Fv]$) is:

```

( .0001  -.0002  .0003 )
(-.0002  .0006  .0000 )
( .0003  .0000  .0029 )

```

The estimated state covariance matrix X is:

```

( .0012  .0001 )
( .0001  .0011 )

```

The relative output error is: $\|y(t) - \bar{y}(t)\| / \|y(t)\| = .65877D-02$

Error Messages:

- 1-
Invalid parameter value on entry.
- 2-
Singular value decomposition failed.
- 3-
A singular upper triangular matrix was found.
- 4-
Not enough working storage. It should be at least //LENG//.

Warnings Messages:

- 1-
More than 100 cycles in sequential data processing
- 2-
All singular values are exactly zero (N = 0)
- 3-
The least squares problem for covariances is rank-deficient
- 4-
The computed output covariance matrix Ry is too small
- 5-
The covariance matrix Ry is numerically singular
- 6-
The transformed matrix A is numerically singular
- 7-
Schur form computation did not converged
- 8-
Schur form ordering failed
- 9-
The computed matrix U_11 is numerically singular

4. Auxiliary Tools for System Identification

To validate and extend the use of various subspace identification schemes, a number of auxiliary routines have been implemented:

- IMMPID estimates a finite set of Markov parameters from input-output data (no constraints on input sequence to be white noise and the initial conditions to be zero).
- IMSHFT shifts the input-output data sequences to compensate the dead-times in the components of the input vector.
- IMICID estimates the initial state $x(0)$ of a linear time-invariant discrete-time system, given the matrix quadruple (A, B, C, D) and the input and output trajectories of the system.
- IMCLID computes the parameters of a linear time-invariant (LTI) system operating in closed-loop with a LTI controller.
- RPIMIU computes the output trajectory of a linear time-invariant discrete-time system, given the input trajectory and/or the initial state and output disturbances (or noise) trajectories.

SUBROUTINE IMMPID

Estimation of Markov Parameters from Input-Output Data

Procedure purpose:

This subroutine determines the order of a linear time-invariant discrete-time state space model and estimates a finite set of Markov parameters using the "ordinary MOESP algorithm with past inputs and outputs".

The model structure is :

$$x(k+1) = Ax(k) + Bu(k) + w(k)$$

$$y(k) = Cx(k) + Du(k) + e(k)$$

where $w(k)$ and $e(k)$ are zero-mean white noise sequences, independent of the noise-free input $u(k)$.

Matrix D could be zero.

The set of Markov parameters is returned as

$$\begin{pmatrix} M_{11} & M_{12} & \dots & M_{1m} \end{pmatrix}$$

$$\begin{pmatrix} M_{21} & M_{22} & \dots & M_{2m} \end{pmatrix}$$

$$\begin{pmatrix} : & : & \dots & : \end{pmatrix}$$

$$\begin{pmatrix} M_{s1} & M_{s2} & \dots & M_{sm} \end{pmatrix}$$

where $M_{1:s,j}$ are $s \times 1$ matrices representing the impulse response of the system due to an impulse in the j -th input.

Usage:

```
CALL IMMPID( NOBL, NSMP, M, L, U, LDU, Y, LDY, N, PARM,
             LDPARM, R, LDR, IWORK, RWORK, LRWORK, TOL,
             CONCT, LAST, CTRL, IWARN, IERR )
```

NOBL : IN, INTEGER

number of block rows in the processed input and output Hankel matrices, s .

NOBL should be chosen larger than n , the estimated dimension of state vector.

NSMP : IN, INTEGER

number of rows of matrices U and Y (number of samples). When sequential data processing is used, NSMP is the number of samples of the current data batch.

NSMP $\geq 2*(M+L+1)*NOBL - 1$, for non-sequential processing;

NSMP $\geq 2*NOBL$, for sequential processing;

NSMP = 0, and LAST = .TRUE., for computing the system matrices only, when the upper triangular factor of the QR factorization is already available in the array R .

The total number of samples when calling the routine with

LAST = .TRUE. should be at least $2*(M+L+1)*NOBL - 1$.

M : IN, INTEGER
 dimension of system input vector.

L : IN, INTEGER
 dimension of system output vector.

U : IN, DOUBLE (LDU,M)
 the NSMP*M input-data sequence matrix U,
 $U = [u_1 \ u_2 \ \dots \ u_m]$. Column j of U contains the
 NSMP values of the j-th input component for consecutive
 time increments.

LDU : IN, INTEGER.
 leading dimension of the array U.
 LDU \geq NSMP.

Y : IN, DOUBLE (LDY,L)
 the NSMP*L system output-data sequence matrix Y,
 $Y = [y_1 \ y_2 \ \dots \ y_l]$. Column j of Y contains the NSMP
 values of the j-th output component for consecutive time
 increments.

LDY : IN, INTEGER.
 leading dimension of the array Y.
 LDY \geq NSMP.

N : OUT, INTEGER
 the order of the system.

PARM : OUT, DOUBLE (LDPARM,L*M)
 If IERR = 0 and IWARN \neq 2, the leading NOBL x L*M
 part of this array contains the estimated Markov
 parameters. The first L columns represent the impulse
 response of the system due to an impulse in the first input;
 the next L columns represent the impulse response of the
 system due to an impulse in the second input and so on.

LDPARM: IN, INTEGER
 leading dimension of the array PARM.
 LDPARM \geq NOBL.

R : OUT, DOUBLE (LDR,2*(M+L)*NOBL)
 If LAST = .FALSE., the leading $2*(M+L)*NOBL \times 2*(M+L)*NOBL$
 part of this array contains the current upper triangular
 factor of the QR factorization used to compress the data
 before determining the singular values in the MOESP algorithm.
 The content of R should be preserved between successive calls
 of the routine with LAST = .FALSE., and the final call with
 LAST = .TRUE..
 During the computations with LAST = .TRUE., the leading
 $M*NOBL \times (M+L)*NOBL$ part of R contains some relevant
 data for the MOESP algorithm (the concatenated compressed
 matrices $R_{11} = R_{1C}'$, and $R_{12} = R_{2C}'$).

LDR : IN, INTEGER
 leading dimension of the array R.

$LDR \geq \max(2*(M+L)*NOBL, 3*M*NOBL)$.
IWORK : OUT, INTEGER(M)
 working array.
RWORK : OUT, DOUBLE (LRWORK)
 On exit, if `LAST = .TRUE.`, and `IERR = 0`:
RWORK(1) contains the optimal suggested value of `LRWORK`;
RWORK(2) contains an estimate of the reciprocal condition
 number for the linear algebraic system solved,
 if `IWARN <> 2`;
RWORK(3:2+L*NOBL) contain singular values used for
 determining the system order.
 The first $(M+L)*(2*NOBL - 1)$ elements of **RWORK** should
 be preserved between successive calls of the routine with
`LAST = .FALSE.`, and the final call with `LAST = .TRUE.`,
 if successive data batches belong to a single experiment
 (`CONCT = .TRUE.`).
 On exit, if `IERR = -1`,
RWORK(1) contains the minimal (estimated) value of `LRWORK`.
LRWORK : IN, INTEGER.
 dimension of working array **RWORK**.
 The value of `LRWORK` must be at least `LRWMIN`, where

$$LRWMIN = \max(4*(NOBL+1)*(M+L)*NOBL, 2+L*NOBL +$$

$$\max((L*NOBL-1)*M*NOBL, 4*(L*NOBL+M),$$

$$5*(L*NOBL+M)-4))$$

 For `NSMP = 0`, the first term of the first `max`
 expression could be omitted.
 This is an overestimate computed using the largest and
 smallest values for `n`, namely `NOBL-1` and `1`, respectively
 (for positive and negative terms, respectively).
 For good performance, `LRWORK` should be large.
TOL : IN, DOUBLE PRECISION.
 absolute tolerance used for determining an estimate of
 the system order. If `TOL >= 0`, the estimate is indicated
 by the index of the last singular value greater than or equal
 to `TOL`. (Singular values less than `TOL` are considered as
 zero.) If `TOL = 0`, an internally computed default value,
 $TOL = NOBL*EPS*SV[1]$, is used, where `EPS` is the relative
 machine precision, and `SV[1]` is the maximal singular value.
 If `TOL < 0`, the estimate is indicated by the index of the
 singular value that has the largest logarithmic gap to its
 successor.
CONCT : IN, LOGICAL
 specifies whether or not the successive data batches in
 sequential data processing belong to a single experiment:
`CONCT = .TRUE.` the current data batch is a continuation
 of the previous data batch.
`CONCT = .FALSE.` there is no connection between the current

data batch and the previous ones.

- LAST : IN, LOGICAL
 specifies whether or not sequential data processing is used, and (for sequential processing) whether or not the current data block is the last one:
 LAST = .TRUE. non-sequential processing or the last data block in sequential processing.
 LAST = .FALSE. not the last data block in sequential processing. In this case the only calculation performed is to update the triangular factor of the QR factorization.
- CTRL : IN, LOGICAL
 specifies whether or not the user's confirmation of the system order estimate is desired, before the computation of system matrices, as follows:
 CTRL = .TRUE. means user's confirmation.
 CTRL = .FALSE. means no confirmation is necessary.
 If CTRL = .TRUE., a reverse communication routine, IMCORD, is called, and, after inspecting the singular values and system order estimate, n, the user may accept n or set a new value.
 IMCORD is not called by the routine if CTRL = .FALSE..
 CTRL = .TRUE. should be used in extreme cases only.
- IWARN : OUT, INTEGER
 unless the routine generates a warning, IWARN contains 0 on exit.
 = 1 : The number of 100 cycles in sequential data processing has been exhausted without signaling that the last block of data was got.
 = 2 : All singular values were exactly zero, hence $N = 0$.
 (Both input and output were identically zero.)
 The Markov parameters could not be determined.
- IERR : OUT, INTEGER
 unless the routine detects an error, IERR contains 0 on exit.
 = 1 : On entry, $NOBL < 1$ or $M < 1$ or $L < 1$ or
 $LDU < NSMP$ or $LDY < NSMP$ or $NSMP < 0$ or
 $NSMP = 0$ (and $LAST = .FALSE.$) or
 $NSMP < 2*(M+L+1)*NOBL-1$ (and $NSMP > 0$ and one data
 block and $LAST = .TRUE.$) or
 $NSMP < 2*NOBL$ (and $LAST = .FALSE.$) or
 $SUM(NSMP) < 2*(M+L+1)*NOBL-1$ (and $LAST = .TRUE.$) or
 $LDR < \max(2*(M+L)*NOBL, 3*M*NOBL)$ or
 $LDPARM < NOBL$.
 = -1 : On entry, $LRWORK < \max(4*(NOBL+1)*(M+L)*NOBL,$
 $2+L*NOBL +$
 $\max((L*NOBL-1)*M*NOBL, 4*(L*NOBL+M),$

$$5*(L*NOBL+M)-4))$$

(For NSMP = 0, the first term of the first max expression could be omitted.)

The minimal value of LRWORK is returned in RWORK(1).

= 2 : The singular value decomposition (SVD) algorithm did not converged.

= 3 : A singular upper triangular matrix was found.

File input/ output:

none

Method:

Let us denote $m = M$, $l = L$, $t = NSPM$, $s = NOBL$.

For non-sequential data processing, the $t \times 2(m+1)s$ matrix

$$\begin{bmatrix} U_f' & U_p' & Y' \\ s+1,s,t & 1,s,t & 1,2s,t \end{bmatrix}$$

is constructed, where U_p , U_f and Y are $1,s,t$, $s+1,s,t$ and $1,2s,t$

Hankel matrices defined in terms of the input and output data [3].

A QR factorization is used to compress the data, and then a singular value decomposition (SVD) of the submatrix

$[R_{24}' \ R_{34}']'$:= $R(ms+1:(2m+1)s, (2m+1)s+1:2(m+1)s)$ of the upper triangular factor R reveals the order n of the system as the number of "non-zero" singular values. Markov parameters are finally computed using the right singular vectors of $[R_{24}' \ R_{34}']'$ and the final, "compressed" submatrices R_{11} and R_{12} of R .

For sequential data processing, the QR decomposition is done sequentially, by updating the upper triangular factor R . When all data have been compressed, the system order and Markov parameters are computed as in the previous case.

Literature

/1/ Verhaegen M., and P. Dewilde

Subspace Model Identification. Part 1: The output-error state-space model identification class of algorithms.

Int. J. Control, 56, pp. 1187-1210, 1992.

/2/ Verhaegen M.

Subspace Model Identification. Part 3: Analysis of the ordinary output-error state-space model identification algorithm.

Int. J. Control, 58, pp. 555-586, 1993.

/3/ Verhaegen M.

Identification of the deterministic part of MIMO state space models given in innovations form from input-output data. Automatica, Vol.30, No.1, pp.61-74, 1994.

Remarks:

- The NSMP argument may vary from a cycle to another in sequential data processing, but NOBL, M, and L should be kept constant. For efficiency, it is advisable to use NSMP as large as possible.
- When 100 cycles of sequential data processing are completed with LAST = .FALSE., a warning is issued, to prevent for an infinite loop.

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Life cycle:

1994 December V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

Consider the original discrete-time system with the following matrices:

$$A = \begin{pmatrix} 1.5 & -0.7 \\ 1.0 & 0.0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0.5 \end{pmatrix}$$

whose output response $y(t)$ to $(0,1)$ random input signals $u(t)$ and zero initial state is determined by using the subroutine RPIMIU. The following sequence of statements can be used to estimate the order and the Markov parameters of the original system:

```

DATA  ISEED/ 1, 2, 3, 5 /
NSMP  = 120
NOBL  = 6
M     = 1
L     = 1
TOL   = -.10000D+01
WITHD = .FALSE.
CONCT = .TRUE.
LAST  = .TRUE.
CTRL  = .TRUE.
LDR   = MAX( 2*(M+L)*NOBL, 3*M*NOBL )
LRWORK = 8100

```

```

C   Compute the output response for uniformly distributed
C   input sequence.
X(1)  = 0.0D0
X(2)  = 0.0D0
CALL RPIMIUI( NSMP, 2, M, L, 0, 0, A, B, E, C, D, F, X, U,
*             W, V, ISEED, Y, RWORK, LRWORK, WITHD, 'G',
*             'G', 'R', 'R', 'R', 1, *1111 )

```

```

C   Compute the system order and Markov parameters.
CALL IMMPID( NOBL, NSMP, M, L, U, NSMP, Y, NSMP, N,
*           PARM, NOBL, R, LDR, IWORK, RWORK, LRWORK,
*           TOL, CONCT, LAST, CTRL, IWARN, IERR )

```

Singular values (in descending order) used to estimate the system order is:

```
2.9705D+01  1.3043D+01  0  0  0  0
```

The system order is $n = 2$. The estimated Markov parameters matrix is:

```
PARM = ( 0.000,  1.000, 2.000, 2.300, 2.050, 1.465 )
```

The maximum relative error (for each column of PARM) is 0.12120D-14.

SUBROUTINE IMSHFT

I/O Data Shifting to Compensate Input Vector Dead-Times

Procedure purpose:

This subroutine shifts the input and output data so that a linear time-invariant system with zero dead-time is obtained. It allows the different input components to have different delays. The initial model structure is :

$$x(k+1) = Ax(k) + \sum_{j=1}^m (B_j u_j(k-t_j)), \quad k \geq t_{\max} + 1, \\ t_{\max} := \max(t_j),$$

$$y(k) = Cx(k) + \sum_{j=1}^m (D_j u_j(k-t_j))$$

where $x(k)$ is the n -dimensional state vector (at time k),
 $u(k)$ is the m -dimensional input vector,
 $y(k)$ is the l -dimensional output vector,
and A , B , C , and D are real matrices of appropriate dimensions.
The final model structure is :

$$\bar{x}(k+1) = \bar{A}\bar{x}(k) + \bar{B}\bar{u}(k), \quad \bar{k} \geq 1,$$

$$\bar{y}(k) = \bar{C}\bar{x}(k) + \bar{D}\bar{u}(k),$$

where $\bar{k} = k - t_{\max}$, $\bar{u}_j(1) = u_j(t_{\max} - t_j + 1)$, $j = 1:m$,
 $\bar{y}(1) = y(t_{\max} + 1)$, $\bar{x}(1) = x(t_{\max} + 1)$.

Usage:

CALL IMSHFT(LTM, NSMP, M, L, U, LDU, Y, LDY, IERR)

LTM : IN, INTEGER (M)
contains the dead-times of the system: LTM(j) is the dead-time corresponding to the j -th input component.

NSMP : IN, OUT, INTEGER
On input : number t of given input and output data values.
On output : number of data values that were shifted.
NSMP > 0.

M : IN, INTEGER
dimension of system input vector.

L : IN, INTEGER
dimension of system output vector.

U : IN, OUT, DOUBLE (NSMP,M)
On input : system input data sequence matrix

$U = [u_1 \ u_2 \ \dots \ u_m]$. The j -th column of U contains the NSMP values of the j -th input component for consecutive time increments.

On output : contains the shifted values of the input data sequence.

LDU : IN, INTEGER

leading dimension of array U .

LDU \geq NSMP.

Y : IN, OUT, DOUBLE (NSMP,L)

On input : system output data sequence matrix

$Y = [y_1 \ y_2 \ \dots \ y_l]$. The j -th column of Y contains the NSMP values of the j -th output component for consecutive time increments.

On output : contains the shifted values of the output data sequence.

LDY : IN, INTEGER

leading dimension of array Y .

LDY \geq NSMP.

IERR : OUT, INTEGER

unless the routine detects an error, IERR contains 0 on exit.

= 1 : On entry, NSMP $<$ 1 or M $<$ 1 or L $<$ 1 or

LDU $<$ NSMP or LDY $<$ NSMP.

File input/ output:

none

Method:

The input and output sequences are shifted, taking into account the maximal dead-time of the system.

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Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

none

SUBROUTINE IMICID

Estimation of the Initial Conditions

Procedure purpose:

To estimate the initial state $x(0)$ of a linear time-invariant (LTI) discrete-time system, given the matrix quadruple (A,B,C,D) and the input and output trajectories of the system.

The model structure is :

$$\begin{aligned} x(k+1) &= Ax(k) + Bu(k), \\ y(k) &= Cx(k) + Du(k), \end{aligned} \quad k \geq 0,$$

where $x(k)$ is the n -dimensional state vector (at time k),
 $u(k)$ is the m -dimensional input vector,
 $y(k)$ is the l -dimensional output vector,
and A , B , C and D are real matrices of appropriate dimensions.

Usage:

```
CALL IMICID( N, M, L, A, LDA, B, LDB, C, LDC, D, LDD,
            U, LDU, Y, LDY, XO, IWORK, RWORK, LRWORK,
            WITHD, WITHSV, IWARN, IERR )
```

```
N      : IN, INTEGER
        the number of system states.
M      : IN, INTEGER
        dimension of system input vector.
L      : IN, INTEGER
        dimension of system output vector.
A      : IN, DOUBLE (LDA,N)
        the N*N state matrix A.
LDA    : IN, INTEGER.
        leading dimension of the array A.
        LDA >= N.
B      : IN, DOUBLE (LDB,M)
        the N*M input matrix B.
LDB    : IN, INTEGER.
        leading dimension of the array B.
        LDB >= N.
C      : IN, DOUBLE (LDC,N)
        the L*N output matrix C.
LDC    : IN, INTEGER.
        leading dimension of the array C.
        LDC >= L.
```

D : IN, DOUBLE (LDD,M)
 the L*M feedthrough matrix D.

LDD : IN, INTEGER.
 leading dimension of the array D.
 LDD >= L.

U : IN, DOUBLE (LDU,M)
 the N*M input-data sequence matrix U,
 $U = [u_1 \ u_2 \ \dots \ u_m]$. Column j of U contains the
 N values of the j-th input component for consecutive
 time increments.

LDU : IN, INTEGER.
 leading dimension of the array U.
 LDU >= N.

Y : IN, DOUBLE (LDY,L)
 the N*L system output-data sequence matrix Y,
 $Y = [y_1 \ y_2 \ \dots \ y_l]$. Column j of Y contains the N
 values of the j-th output component for consecutive time
 increments.

LDY : IN, INTEGER.
 leading dimension of the array Y.
 LDY >= N.

XO : OUT, DOUBLE (N)
 estimated initial state of the system $x(0)$.

IWORK : OUT, INTEGER(N)
 working array.

RWORK : OUT, DOUBLE (LRWORK)
 working array.
 On exit, if IERR = 0, then:
 RWORK(1) contains the optimal suggested value of LRWORK;
 RWORK(2) contains an estimate of the reciprocal
 condition number of the triangular factor of
 QR factorization.
 (RWORK(2) is also set if IERR = 2.)

LRWORK: IN, INTEGER.
 dimension of working array RWORK.
 LRWORK >= L*N*(N+1) + 2*N, if WITHSV = .FALSE.;
 LRWORK >= L*N*(N+1) + N + max(4*N, 5*N - 4),
 if WITHSV = .TRUE..
 For good performance, LRWORK should be larger.

WITHD: IN, LOGICAL.
 specifies whether or not a non-zero feedthrough matrix
 D is to be included in the model, as follows:
 WITHD = .TRUE., (a possibly non-zero matrix D should
 be included);
 WITHD = .FALSE., (a zero matrix D is assumed).

WITHSV: IN, LOGICAL
 specifies whether or not singular value decomposition is

requested to solve the least squares problem, as follows:
 WITHSV = .TRUE., (singular value decomposition is requested);
 WITHSV = .FALSE., (singular value decomposition is not
 requested).

The routine may override this option, if appropriate
 (see IWARN).

IWARN : OUT, INTEGER

unless the routine generates a warning, IWARN contains 0
 on exit.

= 1 : SVD was requested, but it is not needed and was not
 used.

= 2 : SVD was not requested, but it is needed and was
 used.

IERR : OUT, INTEGER

unless the routine detects an error, IERR contains 0 on exit.

= 1 : On entry, $N < 1$ or $M < 1$ or $L < 1$
 or $LDA < N$ or $LDC < L$ or $LDB < N$ or $LDD < L$
 or $LDU < N$ or $LDY < N$.

= -1 : On entry, $LRWORK < L*N*(N+1) + 2*N$
 (and WITHSV = .FALSE.) or
 $LRWORK < L*N*(N+1) + N$
 $+ \max(4*N, 5*N - 4)$
 (and WITHSV = .TRUE.).

The minimal value of LRWORK is returned in RWORK(1).

= 2 : SVD was not requested, but the least squares
 problem has not full column rank.
 The upper triangular matrix of QR factorization was
 found exactly or numerically singular, but there is
 not enough workspace to perform the SVD.

= 3 : The SVD algorithm did not converged.

File input/ output:

none

Method:

The output $y_0(k)$ of the system for zero initial state is
 computed for $k = 0, 1, \dots, n-1$ using the given model.
 Then the following least squares problem is solved for $x(0)$

$$\begin{pmatrix} C \\ CA \\ : \\ CA^{(n-1)} \end{pmatrix} x(0) = \begin{pmatrix} y(0)-y_0(0) \\ y(1)-y_0(1) \\ : \\ y(n-1)-y_0(n-1) \end{pmatrix}$$

The QR decomposition of this matrix is computed.
 If its triangular factor is too ill conditioned,
 then singular value decomposition is used.

Literature

/1/ Verhaegen M., and A. Varga

Some Experience with the MOESP Class of Subspace Model

Identification Methods in Identifying the BO105 Helicopter.

Report TR R165-94, DLR Oberpfaffenhofen, Oberpfaffenhofen, 1994.

Copyright:

1994 - DLR Oberpfaffenhofen, Institut für Robotik und Systemdynamik

Life cycle:

1994 December V. Sima, Institut für Informatik, Bukarest: coded

1995 October A. Varga, DLR Oberpfaffenhofen, -Oberpfaffenhofen: revised

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

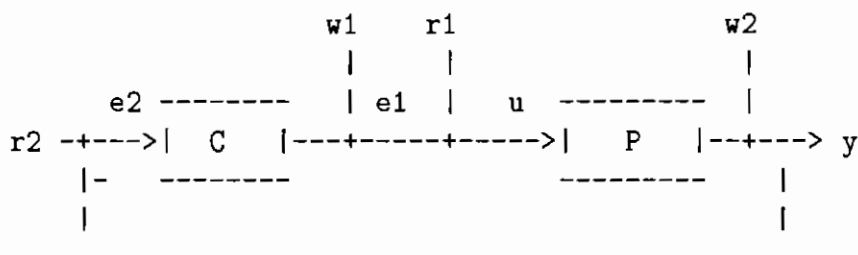
none

SUBROUTINE IMCLID

Closed-Loop Identification of Linear Systems

Procedure purpose:

This subroutine computes non-minimal state space models of a linear time-invariant plant P and/or controller C operating in the closed-loop configuration of the figure below, using a state space model estimate with matrices A, B, C and D, inputs r1 and/or r2 and outputs the signals e1 and/or e2, u and y.



where:

[r1,r2] external excitations;
[w1,w2] external disturbances;
[e1,e2,u,y] internal recordable signals.

The given model of the closed-loop system is $\begin{bmatrix} A & B \\ C & D \end{bmatrix}$

and its signals are:

	inputs	outputs	WITHP	WITHC
(*)	[r1,r2]	[e1,e2,u,y]	.TRUE.	.TRUE.
	[r1]	[e1,u,y] (or *)	.TRUE.	.FALSE.
	[r2]	[e2,u,y] (or *)	.FALSE.	.TRUE.

Usage:

```
CALL IMCLID( N, M, L, ML, LL, A, LDA, B, LDB, C, LDC, D, LDD,
            AP, LDAP, BP, LDBP, CP, LDCP, DP, LDDP, AC, LDAC,
            BC, LDBC, CC, LDCC, DC, LDDC, IWORK, RWORK, LRWORK,
            WITHP, WITHC, IERR )
```

N : IN, INTEGER
order of the system (closed-loop, plant, controller).
M : IN, INTEGER
number of plant inputs.
L : IN, INTEGER
number of plant outputs.
ML : IN, INTEGER
number of closed-loop system inputs.

```

      ML >= M      if WITHP = .TRUE.;
      ML >= L      if WITHC = .TRUE.;
      ML = M + L   if WITHP = .TRUE. and WITHC = .TRUE..
LL   : IN, INTEGER
      number of closed-loop system outputs.
      LL >= 2*M+L  if WITHP = .TRUE.;
      LL >= 2*L+M  if WITHC = .TRUE.;
      LL = 2*(M+L) if WITHP = .TRUE. and WITHC = .TRUE..
A    : IN, DOUBLE (LDA,N)
      the closed-loop system state matrix.
LDA  : IN, INTEGER
      leading dimension of the array A.
      LDA >= N.
B    : IN, DOUBLE (LDB,ML)
      the closed-loop system input matrix.
LDB  : IN, INTEGER
      leading dimension of the array B.
      LDB >= N.
C    : IN, DOUBLE (LDC,N)
      the closed-loop system output matrix.
LDC  : IN, INTEGER
      leading dimension of the array C.
      LDC >= LL.
D    : IN, DOUBLE (LDD,ML)
      the matrix relating the input and output of the
      closed-loop system.
LDD  : IN, INTEGER
      leading dimension of the array D.
      LDD >= LL.
AP   : OUT, DOUBLE (LDAP,N)
      on exit, if IERR = 0, or IERR = 3, AP contains
      the plant state matrix Ap.
      AP is not referenced if WITHP = .FALSE..
LDAP : IN, INTEGER
      leading dimension of the array AP.
      LDAP >= N if WITHP = .TRUE. and LDAP >= 1, otherwise.
BP   : OUT, DOUBLE (LDBP,M)
      on exit, if IERR = 0, or IERR = 3, BP contains the
      plant input matrix Bp.
      BP is not referenced if WITHP = .FALSE..
LDBP : IN, INTEGER
      leading dimension of the array BP.
      LDBP >= N if WITHP = .TRUE. and LDBP >= 1, otherwise.
CP   : OUT, DOUBLE (LDCP,N)
      on exit, if IERR = 0, or IERR = 3, CP contains the
      plant output matrix Cp.
      CP is not referenced if WITHP = .FALSE..

```

LDCP : IN, INTEGER
 leading dimension of the array CP.
 LDCP \geq L if WITHP = .TRUE. and LDCP \geq 1, otherwise.

DP : OUT, DOUBLE (LDDP,M)
 on exit, if IERR = 0, or IERR = 3, DP contains the
 matrix relating the input and output of the plant Dp.
 DP is not referenced if WITHP = .FALSE..

LDDP : IN, INTEGER
 leading dimension of the array DP.
 LDDP \geq L if WITHP = .TRUE. and LDDP \geq 1, otherwise.

AC : OUT, DOUBLE (LDAC,N)
 on exit, if IERR = 0, or IERR = 2, AC contains the
 controller state matrix Ac.
 AC is not referenced if WITHC = .FALSE..

LDAC : IN, INTEGER
 leading dimension of the array AC.
 LDAC \geq N if WITHC = .TRUE. and LDAC \geq 1, otherwise.

BC : OUT, DOUBLE (LDBC,L)
 on exit, if IERR = 0, or IERR = 2, BC contains the
 controller input matrix Bc.
 BC is not referenced if WITHC = .FALSE..

LDBC : IN, INTEGER
 leading dimension of the array BC.
 LDBC \geq N if WITHC = .TRUE. and LDBC \geq 1, otherwise.

CC : OUT, DOUBLE (LDCC,N)
 on exit, if IERR = 0, or IERR = 2, CC contains the
 controller output matrix Cc.
 CC is not referenced if WITHC = .FALSE..

LDCC : IN, INTEGER
 leading dimension of the array CC.
 LDCC \geq M if WITHC = .TRUE. and LDCC \geq 1, otherwise.

DC : OUT, DOUBLE (LDDC,L)
 on exit, if IERR = 0, or IERR = 2, DC contains the
 matrix relating the input and output of the controller Dc.
 DC is not referenced if WITHC = .FALSE..

LDDC : IN, INTEGER
 leading dimension of the array DC.
 LDDC \geq M if WITHC = .TRUE. and LDDC \geq 1, otherwise.

IWORK : OUT, INTEGER(max(M,L))
 working array.

RWORK : OUT, DOUBLE (LRWORK)
 working array.
 RWORK(1) contains an estimate of the reciprocal condition
 number of the matrix D_31 (if WITHP = .TRUE.);
 RWORK(2) contains an estimate of the reciprocal condition
 number of the matrix D_22 (if WITHC = .TRUE.).
 (See Method.)

LRWORK: IN, INTEGER
dimension of working array RWORK.
The value of LRWORK must be at least LRWMIN, where
 $LRWMIN = (N + M + 4)*M$ if WITHP = .TRUE.;
 $LRWMIN = (N + L + 4)*L$ if WITHC = .TRUE.;
 $LRWMIN = \max((N+M+4)*M, (N+L+4)*L)$
if WITHP = .TRUE. and WITHC = .TRUE..

WITHP : IN, LOGICAL
specifies whether or not the plant model is to be computed,
as follows:
WITHP = .TRUE., (the plant model should be computed);
WITHP = .FALSE., (the plant model should not be computed).

WITHC : IN, LOGICAL
specifies whether or not the controller model is to be
computed, as follows:
WITHC = .TRUE., (the controller model should be
computed);
WITHC = .FALSE., (the controller model should not be
computed).

IERR : OUT, INTEGER
Unless the routine detects an error, IERR contains 0 on exit.

= 1 : On entry, $N < 1$ or $M < 1$ or $L < 1$ or
 $ML < M$ (and WITHP = .TRUE.) or
 $ML < L$ (and WITHC = .TRUE.) or
 $ML \neq M + L$ (and WITHP = .TRUE. and WITHC = .TRUE.) or
 $LL < 2*M+L$ (and WITHP = .TRUE.) or
 $LL < 2*L+M$ (and WITHC = .TRUE.) or
 $LL \neq 2*(M+L)$ (and WITHP = .TRUE. and WITHC = .TRUE.)
or $LDA < N$ or $LDB < N$ or $LDC < LL$ or $LDD < LL$ or
or $LDAP < N$ or $LDBP < N$ or $LDCP < L$ or $LDDP < L$
(and WITHP = .TRUE.)
or $LDAP < 1$ or $LDBP < 1$ or $LDCP < 1$ or $LDDP < 1$
(and WITHP = .FALSE.)
or $LDAC < N$ or $LDBC < N$ or $LDCC < M$ or $LDDC < M$ or
(and WITHC = .TRUE.)
or $LDAC < 1$ or $LDBC < 1$ or $LDCC < 1$ or $LDDC < 1$
(and WITHC = .FALSE.).

= -1 : On entry,
 $LRWORK < (N + M + 4)*M$ if WITHP = .TRUE.;
 $LRWORK < (N + L + 4)*L$ if WITHC = .TRUE.;
 $LRWORK < \max((N+M+4)*M, (N+L+4)*L)$
if WITHP = .TRUE. and WITHC = .TRUE..
The minimal value of LRWORK is returned in
RWORK(1).

= 2 : The matrix D_31 is exactly singular.
The plant matrices could not be computed.

= 3 : The matrix D_22 is exactly singular.

The controller matrices could not be computed.

File input/ output:

none

Method:

The computing formulas for the plant model are as follows:

$$\begin{array}{l}
 \text{Ap} = \text{A} - \text{B}_1 \text{D}_{31}^{-1} \text{C}_3, \quad \text{Bp} = \text{B}_1 \text{D}_{31}^{-1}, \\
 \text{Cp} = \text{C}_4 - \text{D}_{41} \text{D}_{31}^{-1} \text{C}_3, \quad \text{Dp} = \text{D}_{41} \text{D}_{31}^{-1},
 \end{array}$$

where B_1 , C_3 , C_4 , D_{31} , and D_{41} are defined in terms of A , B , C , D /1/. Similar relations hold for the controller model.

Literature

/1/ Verhaegen M.

Application of a Subspace Model Identification technique
to identify LTI systems operating in closed-loop.
Automatica, 29, pp. 1027-1040, 1993.

Copyright:

1994 - DLR Oberpfaffenhofen, Institut für Robotik und Systemdynamik

Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

none

SUBROUTINE RPIMIU

Simulation of Linear Time-Invariant Discrete-Time Systems

Procedure purpose:

This subroutine computes the output trajectory of a linear time-invariant discrete-time system, given the input trajectory and/or the initial state and output disturbances (or noise) trajectories.

The model structure is :

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) + Ew(k), \quad k \geq 1, \\y(k) &= Cx(k) + Du(k) + Fv(k),\end{aligned}$$

where $x(k)$ is the n -dimensional state vector (at time k),
 $u(k)$ is the m -dimensional input vector,
 $y(k)$ is the l -dimensional output vector,
 $w(k)$ is the mw -dimensional state disturbance vector,
 $v(k)$ is the lv -dimensional output disturbance vector,
and $A, B, C, D, E,$ and F are real matrices of appropriate dimensions. The matrix D is taken as zero if $WITHD = .FALSE..$

Usage:

```
CALL RPIMIU( NSMP, N, M, L, MW, LV, A, B, E, C, D, F,
            X, U, W, V, ISEED, Y, RWORK, LRWORK, WITHD,
            UNITE, UNITF, UTRJ, WTRJ, VTRJ, IDIST, * )
```

```
NSMP : IN, INTEGER
      number of time steps where y should be computed.
      NSMP > 0.
N     : IN, INTEGER
      number of system states.
M     : IN, INTEGER
      dimension of system input vector.
      M >= 0.
L     : IN, INTEGER
      dimension of system output vector.
MW    : IN, INTEGER
      number of state disturbances.
      MW >= 0.
LV    : IN, INTEGER
      number of output disturbances.
      LV >= 0.
A     : IN, DOUBLE (N,N)
      the N*N system state matrix A.
      (column dense)
```

B : IN, DOUBLE (N,M)
 the N*M system input matrix B.
 B is not referenced if M = 0.
 (column dense)

E : IN, DOUBLE (N,MW)
 the N*MW state disturbance matrix E.
 E is not referenced if MW = 0 or UNITE = 'I'.
 (column dense)

C : IN, DOUBLE (L,N)
 the L*N output matrix C.
 (column dense)

D : IN, DOUBLE (L,M)
 the L*M input-output matrix D if WITHD = .TRUE..
 D is not referenced if M = 0 or WITHD = .FALSE..
 (column dense)

F : IN, DOUBLE (L,LV)
 the L*LV output disturbance matrix F.
 F is not referenced if LV = 0 or UNITF = 'I'.

X : IN, OUT, DOUBLE (N)
 On input : the initial state of the system X,
 $X = [x_1 \ x_2 \ \dots \ x_n]$.
 On output : the final state of the system.
 The state trajectory is computed, but not stored.

U : IN, OUT, DOUBLE (NSMP,M)
 On input : the NSMP*M input data sequence matrix U,
 $U = [u_1 \ u_2 \ \dots \ u_m]$.
 If UTRJ = 'G', the j-th column of U must contain
 the NSMP values of the j-th input component for
 consecutive time increments.
 If UTRJ = 'I', or UTRJ = 'S', only the first row
 of U, containing the values for the impulse or
 step signal applied to the system, must be given.
 U should not be given if UTRJ = 'R'.
 U is not referenced if M = 0.
 On output : if M > 0 and UTRJ = 'R', U contains the random
 input trajectory; otherwise U is unchanged.
 (column dense)

W : IN, OUT, DOUBLE (NSMP,MW)
 On input : the NSMP*MW state disturbance (noise) data sequence
 matrix W, $W = [w_1 \ w_2 \ \dots \ w_{mw}]$.
 If WTRJ = 'G', the j-th column of W contains the
 NSMP values of the j-th state disturbance component
 for consecutive time increments.
 If WTRJ = 'I', or WTRJ = 'S', only the first row
 of W, containing the values for the impulse or
 step signal applied to the system, must be given.
 W should not be given if WTRJ = 'R'.

W is not referenced if MW = 0.

On output : if MW > 0 and WTRJ = 'R', the first row of W contains the last randomly generated state disturbance vector; otherwise W is unchanged.

(column dense)

V : IN, OUT, DOUBLE (NSMP,LV)

On input : the NSMP*LV output disturbance (noise) data sequence matrix V, $V = [v_1 \ v_2 \ \dots \ v_{lv}]$.
 If VTRJ = 'G', the j-th column of V contains the NSMP values of the j-th output disturbance component for consecutive time increments.
 If VTRJ = 'I', or VTRJ = 'S', only the first row of V, containing the values for the impulse or step signal applied to the system, must be given.
 V should not be given if VTRJ = 'R'.
 V is not referenced if LV = 0.

On output : if LV > 0 and VTRJ = 'R', the first row of V contains the last randomly generated output disturbance vector; otherwise V is unchanged.

(column dense)

ISEED : IN, OUT, INTEGER(4)

On input : If either UTRJ = 'R', or WTRJ = 'R', or VTRJ = 'R', ISEED contain the seed of the random number generator; the array elements must be between 0 and 4095, and ISEED(4) must be odd. The random multidimensional deviates are computed at each time step, in the order: V, U, and W. ISEED is not referenced if neither UTRJ, nor WTRJ, nor VTRJ is set to 'R'.

On output : If either UTRJ = 'R', or WTRJ = 'R', or VTRJ = 'R', ISEED contain the final updated seed of the random number generator.

Y : OUT, DOUBLE (NSMP,L)

the NSMP*L system output data sequence matrix Y, $Y = [y_1 \ y_2 \ \dots \ y_l]$. The j-th column of Y contains the NSMP values of the j-th output component for consecutive time increments.

(column dense)

RWORK : OUT, DOUBLE (LRWORK)

working array.

LRWORK: IN, INTEGER

dimension of working array RWORK.
 The value of LRWORK must be at least LRWMIN, where $LRWMIN = \max(L,N) + M$ if UTRJ = 'R'.
 $LRWMIN \geq \max(L,N)$, otherwise.

WITHD : IN, LOGICAL

specifies whether or not D is a non-zero matrix to be

included in the model:
 WITHD = .TRUE., D is a non-zero matrix and is used;
 WITHD = .FALSE., D is a zero matrix.

UNITE : IN, CHARACTER*1
 specifies whether or not E is the identity matrix:
 UNITE = 'I', E is the identity matrix of order n;
 In this case MW should be equal to N.
 UNITE = 'G', E is a general matrix.

UNITF : IN, CHARACTER*1
 specifies whether or not F is the identity matrix:
 UNITF = 'I', F is the identity matrix of order l;
 In this case LV should be equal to L.
 UNITF = 'G', F is a general matrix.

UTRJ : IN, CHARACTER*1
 specifies the desired time trajectory for the input U:
 UTRJ = 'G', U is a given time trajectory;
 UTRJ = 'I', U is an impulse signal applied on the first k;
 UTRJ = 'S', U is a step signal;
 UTRJ = 'R', U is a random m-dimensional deviate.

WTRJ : IN, CHARACTER*1
 specifies the desired time trajectory for the state
 disturbance W:
 WTRJ = 'G', W is a given time trajectory;
 WTRJ = 'I', W is an impulse signal applied on the first k;
 WTRJ = 'S', W is a step signal;
 WTRJ = 'R', W is a random mw-dimensional deviate.

VTRJ : IN, CHARACTER*1
 specifies the desired time trajectory for the output
 disturbance V:
 VTRJ = 'G', V is a given time trajectory;
 VTRJ = 'I', V is an impulse signal applied on the first k;
 VTRJ = 'S', V is a step signal;
 VTRJ = 'R', V is a random lv-dimensional deviate.

IDIST : IN, INTEGER.
 if either UTRJ = 'R', or WTRJ = 'R', or VTRJ = 'R',
 IDIST specifies the distribution of the random numbers:
 IDIST = 1, uniform (0,1);
 IDIST = 2, uniform (-1,1);
 IDIST = 3, normal (0,1).
 Otherwise IDIST is not referenced.

* : RETURN 1, target label in case of error (e.g. *1111)

File input/ output:

none

Method:

For the specified values of the input vector u, state disturbance

vector w and output disturbance vector v , the output of the system is computed step by step, using the given system model:

$$\begin{aligned}x(k+1) &= Ax(k) + Bu(k) + Ew(k) \\y(k) &= Cx(k) + Du(k) + Fv(k).\end{aligned}$$

For `WITHD = .FALSE.`, the matrix D is taken as zero.

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Life cycle:

1994 MARCH V. Sima, Institut für Informatik, Bukarest: coded

Libraries required:

RASP, BLAS (1,2,3), LAPACK

Example:

none

Error Messages:

-1-

Invalid parameter value on entry.

Appendix 1. RASP-IDENT Driver Routines

Tools for State Intersection Class of Subspace Model Identification

RPIMN4 computes, by using the basic N4SID algorithm, consistent and statistically efficient estimates of the matrices (A,B,C,D) of a state space model when the output is additively perturbed by a zero-mean white noise independent from the input, and the disturbance input is generated by an innovation model. Optionally, also calculates the noise covariances and the Kalman gain to allow the design of a state observer. RASP interface to the subroutine IMN4S.

Tools for MOESP Class of Subspace Model Identification

RPIMOE computes consistent and statistically efficient estimates of the matrices (A,B,C,D) of a state space model when the output is additively perturbed by a zero-mean white noise independent from the input. The ordinary MOESP scheme is used. RASP interface to the subroutine IMOESO.

RPIMPI computes consistent estimates of the matrices (A,B,C,D) of a state space model when the output is additively perturbed by a zero-mean but of arbitrary statistical color. The MOESP scheme extended with instrumental variables based on past input quantities is used. RASP interface to the subroutine IMOEPI.

RPIMPO computes consistent and statistically efficient estimates of the matrices (A,B,C,D) of a state space model when the output is additively perturbed by a zero-mean white noise independent from the input, and the disturbance input is generated by an innovation model. The MOESP scheme extended with instrumental variables based on past input and output quantities is used. RASP interface to the subroutine IMOEPO.

RPIMRS computes consistent estimates of the matrices (A,B,C,D) of a state space model when the output is additively perturbed by a zero-mean, but of arbitrary statistical color. The MOESP scheme extended with instrumental variables based on reconstructed state and/or past input quantities is used. RASP interface to the subroutine IMOERS.

RPIMKG in addition to the RPIMPO routine, also calculates the noise covariances and the corresponding Kalman gain to allow the design of a state observer. RASP interface to the subroutine IMOEKG.

Auxiliary Tools for System Identification

- IMMPID estimates a finite set of Markov parameters from input-output data (no constraints on input sequence to be white noise and the initial conditions to be zero).
- IMSHFT shifts the input-output data sequences to compensate the dead-times in the components of the input vector.
- IMICID estimates the initial state $x(0)$ of a linear time-invariant discrete-time system, given the matrix quadruple (A, B, C, D) and the input and output trajectories of the system.
- IMCLID computes the parameters of a linear time-invariant (LTI) system operating in closed-loop with a LTI controller.
- RPIMIU computes the output trajectory of a linear time-invariant discrete-time system, given the input trajectory and/or the initial state and output disturbances (or noise).
RASP-IDENT interface to the subroutine IMOUTP.

Appendix 2. SLICOT Compatible Driver Routines

- IMCLID computes the parameters of a linear time-invariant system operating in closed-loop with a linear time-invariant controller.
- IMMPID estimates a finite set of Markov parameters from input-output data (no constraints on input sequence to be white noise and the initial conditions to be zero).
- IMN4S determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices, the noise covariance matrices and the Kalman gain matrix using the basic N4SID algorithm.
- IMOEBG determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices, the noise covariance matrices and the Kalman gain matrix using the MOESP algorithm with past inputs and outputs (PO) combined with the N4SID approach.
- IMOEBI determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices using the "ordinary MOESP algorithm with past inputs".
- IMOEBP determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices using the "ordinary MOESP algorithm with past inputs and outputs".
- IMOERS determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices using the MOESP algorithm with reconstructed state variables as instrumental variables.
- IMOESO determines the order of a linear time-invariant discrete-time state space model and estimates the quadruple of system matrices using the "ordinary MOESP algorithm".
- IMOUTP computes the output trajectory of a linear time-invariant discrete-time system, given the input trajectory and/or the initial state and output disturbances (or noise).
- IMSHFT shifts the input and output data so that a linear time-invariant system with dead-time zero is obtained. It allows the different input quantities to have different delays.

Appendix 3. SLICOT Compatible Computational and Auxiliary Routines

- DGELSR computes the rank-revealing QR factorization of a real general matrix A , which may be rank-deficient, and applies the left transformations to another matrix B .
- DGELSZ uses the rank-revealing QR factorization of a real general matrix A , computed by DGELSR, to determine the minimum-norm solution to a real linear least squares problem: minimize $\|A * X - B\|$.
- DORQRC overwrites the real $(n+m)*l$ matrix $C = \begin{pmatrix} A \\ B \end{pmatrix}$, where A and B are real $n*l$ and $m*l$ matrices, respectively, with $Q'*C$, where Q is a real orthogonal matrix defined as the product of elementary Householder reflectors as returned by DTRQRC.
- DORQRT overwrites the real matrix C with $Q' * C$, or $Q * C$, where Q is a real orthogonal matrix defined as the product of elementary Householder reflectors as returned by DTRQRC.
- DTRQRC computes a QR factorization, $C = Q * R$, of the real $(n+m)*n$ matrix $C = \begin{pmatrix} A \\ B \end{pmatrix}$, where A is an $n*n$ upper triangular matrix and B is an $m*n$ matrix.
- DTRQRT computes a QR factorization of a real $(m+n)$ by n matrix A ($A = Q * R$), whose trailing n by n submatrix is upper triangular.
- DTRSVB returns part of the singular value decomposition of a real upper triangular matrix and optionally applies the left transformations to another matrix.
- EXTEND constructs the Hamiltonian or symplectic matrix associated to the linear-quadratic optimization problem, used to solve a continuous- or discrete-time algebraic Riccati equation, respectively.
- IMABCD estimates the quadruple of system matrices A , B , C , and D of a linear time-invariant state space model using the singular value decomposition information provided by other routines.

- IMADN4 estimates the quadruple of system matrices A, B, C, and D, and, optionally, the noise covariance matrices of the linear time-invariant state space model using the N4SID approach.
- IMBDN4 estimates the system matrices B and D of the linear time-invariant state space model for the N4SID approach.
(auxiliary routine called only by IMADN4 for efficient use of the memory space)
- IMCORD asks for user's confirmation of the system order found by other subroutines. This reverse communication routine could be modified, but its interface must be preserved.
- IMKPRE determines the Kalman gain (predictor) matrix of a linear time-invariant discrete-time state space model using the system and noise covariance matrices.
- IMPARM estimates the Markov parameters of a linear time-invariant system using the singular value decomposition information provided by other routines.
(auxiliary routine called only by IMMPID)
- IMTRQ computes the triangular QR factor of a structured matrix.
(auxiliary routine called only by IMABCD for efficient use of the memory space)
- IMTRQN computes the triangular QR factor of a structured matrix.
(auxiliary routine called only by IMBDN4 for efficient use of the memory space)
- OPTREG computes the state feedback gain matrix of continuous- or discrete-time optimal regulator, based on the solution of an algebraic Riccati equation.
- RICSCH computes the solution matrix of continuous- or discrete-time algebraic Riccati equation, or an orthogonal basis for the stable or unstable invariant subspace, using the Schur vectors approach.
The Hamiltonian or symplectic matrix associated to the linear-quadratic optimization problem is assumed to be given.

Appendix 4. Called RASP-MODRED, LAPACK and BLAS Routines

RASP-MODRED Routines

DGEES1, DMNEG, DMTRA, DTRSVD, RCFI, RPARPR, SEOR1, SPLITB

Directly called LAPACK Routines

DBDSQR, DGEBRD, DGECON, DGEQPF, DGEQRF, DGESV, DGETRF, DGETRI, DGETRS,
DORGBR, DORM2R, DORMBR, DORMQR, DPOCON, DPOTRF, DPOTRS, DTRCON, DTRTRI,
DTRTRS, DTZRQF

DLABAD, DLACPY, DLAIC1, DLANGE, DLANSY, DLANTR, DLARF, DLARFG, DLARNV,
DLASCL, DLASET, DLATZM

DLAMCH, ILAENV, LSAME

Directly called BLAS 1 Routines

DAXPY, DCOPY, DDOT, DNRM2, DSCAL, DSWAP

Directly called BLAS 2 Routines

DGEMV, DGER, DTRMV, XERBLA

Directly called BLAS 3 Routines

DGEMM, DSYMM, DSYRK, DTRSM

RASP Error Processing Routines

RPERRB, RPERRI, RPERRN, RPERRP, RPERRT, RPERRV, RPINIT, RPQSLE