

Work Package 4B: Computing the global minimum of least squares system identification and H_2 model reduction problems



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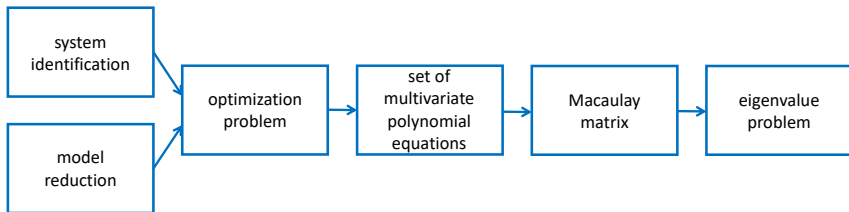


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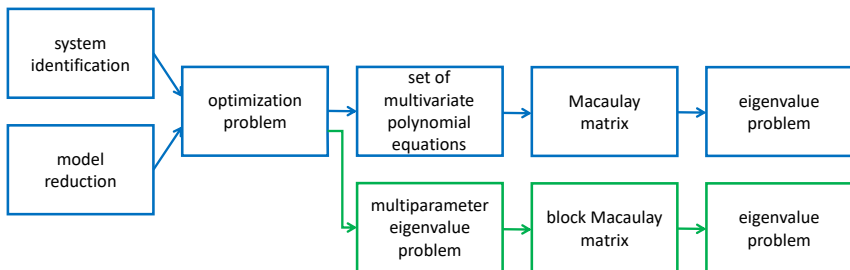
Strategy

- presented at Selma retreat in Mons, January 2019



Strategy

- presented at Selma retreat in Mons, January 2019
- improved during Selma project¹



¹Bart De Moor, Least squares optimal realisation of autonomous LTI systems is an eigenvalue problem, Communications in Information and systems, vol. 20, no. 2, pp. 163–207, 2020

Christof Vermeersch and Bart De Moor, Globally optimal least-squares ARMA model identification is an eigenvalue problem, IEEE Control Systems Letters (L-CSS), vol. 2, no. 4, pp. 1062–1067, 2019

Overview

- The multiparameter eigenvalue problem (MEP)
- Solving an MEP using a block Macaulay matrix: simple example
- MEPs from identification problems
- Recent progress
 1. Algorithms
 2. Theory

The multiparameter eigenvalue problem (MEP)

- Standard eigenvalue problem (SEP)

$$(A - \lambda I) v = 0$$

- Generalized eigenvalue problem (GEP)

$$(A - \lambda B) v = 0$$

- Polynomial eigenvalue problem (PEP)

$$(P_0 + P_1\lambda + \cdots + P_k\lambda^k) v = 0$$

- Multiparameter eigenvalue problem (MEP)

example: 3-parameter $(\lambda_1, \lambda_2, \lambda_3)$ eigenvalue problem of degree 4

$$(P_{000} + P_{100}\lambda_1 + P_{002}\lambda_3^2 + P_{013}\lambda_2\lambda_3^3) v = 0$$

Solving an MEP using a block Macaulay matrix

- working example

$$(P_{00} + P_{10}\lambda + P_{01}\mu)v = 0$$

$$P_{00} = \begin{pmatrix} 2 & -5 \\ -2 & -1 \\ 5 & -1 \end{pmatrix}, P_{10} = \begin{pmatrix} 3 & 0 \\ 3 & -1 \\ -3 & 2 \end{pmatrix}, P_{01} = \begin{pmatrix} 2 & 2 \\ 3 & 2 \\ -2 & -4 \end{pmatrix}$$

find the eigenvalue tuples (λ, μ) and eigenvectors v

- three solutions (one real, two complex, no solutions at infinity)

$$\begin{array}{l} \lambda_1 = 3.4536 \\ \lambda_2 = -0.2268 + 1.4608i \\ \lambda_3 = -0.2268 - 1.4608i \end{array} \left| \begin{array}{l} \mu_1 = 1.1169 \\ \mu_2 = 0.4415 - 0.7775i \\ \mu_3 = 0.4415 + 0.7775i \end{array} \right| \begin{array}{l} v_1 = \begin{pmatrix} 0.1862 \\ 0.9825 \end{pmatrix} \\ v_2 = \begin{pmatrix} -0.4946 + 0.5971i \\ -0.5972 + 0.2053i \end{pmatrix} \\ v_3 = \begin{pmatrix} -0.4946 - 0.5971i \\ -0.5972 - 0.2053i \end{pmatrix} \end{array}$$

- create extra equations

$$\begin{aligned} (P_{00} + P_{10}\lambda + P_{01}\mu)v &= 0 \\ \lambda(P_{00} + P_{10}\lambda + P_{01}\mu)v &= 0 \\ \mu(P_{00} + P_{10}\lambda + P_{01}\mu)v &= 0 \end{aligned}$$

- result: block-Macaulay matrix \mathcal{M}

$$\underbrace{\begin{pmatrix} P_{00} & P_{10} & P_{01} & 0 & 0 & 0 \\ 0 & P_{00} & 0 & P_{10} & P_{01} & 0 \\ 0 & 0 & P_{00} & 0 & P_{10} & P_{01} \end{pmatrix}}_{\mathcal{M}} \begin{pmatrix} v \\ \lambda v \\ \mu v \\ \lambda^2 v \\ \lambda \mu v \\ \mu^2 v \end{pmatrix} = 0$$

- null space of \mathcal{M} has dimension 3
- vectors k_1 , k_2 and k_3 span the null space of \mathcal{M}

$$k_1 = \begin{pmatrix} v_1 \\ \lambda_1 v_1 \\ \mu_1 v_1 \\ \lambda_1^2 v_1 \\ \lambda_1 \mu_1 v_1 \\ \mu_1^2 v_1 \end{pmatrix} \quad k_2 = \begin{pmatrix} v_2 \\ \lambda_2 v_2 \\ \mu_2 v_2 \\ \lambda_2^2 v_2 \\ \lambda_2 \mu_2 v_2 \\ \mu_2^2 v_2 \end{pmatrix} \quad k_3 = \begin{pmatrix} v_3 \\ \lambda_3 v_3 \\ \mu_3 v_3 \\ \lambda_3^2 v_3 \\ \lambda_3 \mu_3 v_3 \\ \mu_3^2 v_3 \end{pmatrix}$$

- calculate a basis for null space of \mathcal{M} : columns of Γ

- columns of Γ are linear combinations of vectors k_1, k_2, k_3 :

$$\Gamma_{\text{calculated}} = \underbrace{\begin{pmatrix} v_1 & v_2 & v_3 \\ \lambda_1 v_1 & \lambda_2 v_2 & \lambda_3 v_3 \\ \mu_1 v_1 & \mu_2 v_2 & \mu_3 v_3 \\ \lambda_1^2 v_1 & \lambda_2^2 v_2 & \lambda_3^2 v_3 \\ \lambda_1 \mu_1 v_1 & \lambda_2 \mu_2 v_2 & \lambda_3 \mu_3 v_3 \\ \mu_1^2 v_1 & \mu_2^2 v_2 & \mu_3^2 v_3 \end{pmatrix}}_{\text{unknown}} T \quad (T \text{ is nonsingular})$$

- Γ has special structure

$$\Gamma = \begin{pmatrix} C \\ CA_1 \\ CA_2 \\ CA_1^2 \\ CA_1 A_2 \\ CA_2^2 \end{pmatrix} \quad \text{where} \quad \begin{cases} C_{2 \times 3} = (v_1 \ v_2 \ v_3) T \\ A_1_{3 \times 3} = T^{-1} \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{pmatrix} T \\ A_2_{3 \times 3} = T^{-1} \begin{pmatrix} \mu_1 & & \\ & \mu_2 & \\ & & \mu_3 \end{pmatrix} T \end{cases}$$

- from Γ we can find C : the first block row of Γ
- from Γ we can find A_1 and A_2 by solving linear equations
 - selection of block rows

$$\underbrace{\begin{pmatrix} C \\ CA_1 \\ CA_2 \\ CA_1^2 \\ CA_1A_2 \\ CA_2^2 \end{pmatrix}}_{\Gamma} \xrightarrow{\cdot A_1} \underbrace{\begin{pmatrix} C \\ CA_1 \\ CA_2 \\ CA_1^2 \\ CA_1A_2 \\ CA_2^2 \end{pmatrix}}_{\Gamma}$$

$$\underbrace{\begin{pmatrix} C \\ CA_1 \\ CA_2 \\ CA_1^2 \\ CA_1A_2 \\ CA_2^2 \end{pmatrix}}_{\Gamma} \xrightarrow{\cdot A_2} \underbrace{\begin{pmatrix} C \\ CA_1 \\ CA_2 \\ CA_1^2 \\ CA_1A_2 \\ CA_2^2 \end{pmatrix}}_{\Gamma}$$

- solving equations

$$\begin{pmatrix} C \\ CA_1 \\ CA_2 \end{pmatrix} A_1 = \begin{pmatrix} CA_1 \\ CA_1^2 \\ CA_1A_2 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} C \\ CA_1 \\ CA_2 \end{pmatrix} A_2 = \begin{pmatrix} CA_2 \\ CA_1A_2 \\ CA_2^2 \end{pmatrix}$$

- recall:

$$C = (v_1 \quad v_2 \quad v_3) T$$
$$A_1 = T^{-1} \begin{pmatrix} \lambda_1 & & \\ & \lambda_2 & \\ & & \lambda_3 \end{pmatrix} T$$
$$A_2 = T^{-1} \begin{pmatrix} \mu_1 & & \\ & \mu_2 & \\ & & \mu_3 \end{pmatrix} T$$

- the eigenvalues of A_1 and A_2 give the solutions (λ_i, μ_i) ($i = 1, 2, 3$) of the MEP
- the eigenvectors of A_1 are equal to the eigenvectors of A_2 and form the matrix T
- the eigenvectors of the MEP can be found as

$$(v_1 \quad v_2 \quad v_3) = CT^{-1}$$

MEPs from identification problems

- big MEPs
- not only linear terms in the MEP
- solutions at infinity
- solution set at infinity, dimension > 0

example: first order ARMA model with parameters α and β :

$$(P_{00} + P_{10}\alpha + P_{01}\beta + P_{20}\alpha^2 + P_{11}\alpha\beta + P_{02}\beta^2) v = 0$$

size of the matrices: $(3N - 1) \times (3N - 2)$, where N is the number of output measurements

Recent progress: Algorithms

1. recursive and sparse algorithms to compute the null space of the block Macaulay matrix²

- faster
- less memory

example: identification of order 1 ARMA model from $N = 8$ data points

- size of matrices in MEP: 23×22
- size of block Macaulay matrix: $20\,769 \times 21\,780$
- memory usage: **factor 150 000 reduction**
 - standard approach: 3.62 GB
 - sparse adaptation: 24.28 kB
- computation time: **factor 725 faster**
 - standard approach: 30 225 s (8.5 hours)
 - sparse-recursive: 42 s

²Christof Vermeersch and Bart De Moor, Two Double Recursive Block Macaulay Matrix Algorithms to Solve Multiparameter Eigenvalue Problems, accepted

2. avoid computation of null space, work directly in column space of block Macaulay matrix³

³Christof Vermeersch and Bart De Moor, A column space based approach to solve systems of multivariate polynomial equations, IFAC-PapersOnLine, vol. 54, no. 9, pp. 137–144, July 2021

Christof Vermeersch and Bart De Moor, Two Complementary Block Macaulay Algorithms to Solve Multiparameter Eigenvalue Problems, submitted

Recent progress: Theory

1. different types of eigenvalue problems and their connection to shift-invariant spaces and realization theory⁴, see poster [Christof Vermeersch](#)
2. study of multidimensional (mD) systems
 - an mD system has more than one independent variable, e.g., time and place
 - connection with our Selma work:
 Γ of the MEP example is the observability matrix of a 2-dimensional LTI system with commuting system matrices A_1 and A_2 and output matrix C
 - results:
 - 2D descriptor systems: condition for system so that non-trivial state sequence exists⁵
 - observability
 - mD realization: find commutative system from output data

⁴Katrien De Cock and Bart De Moor, Multiparameter Eigenvalue Problems and Shift-Invariance, IFAC PapersOnLine, vol. 54, no. 9, pp. 159–165, July 2021

⁵Bob Vergauwen and Bart De Moor, Two-dimensional descriptor systems, IFAC PapersOnLine vol. 54, no. 9, pp. 151–158, July 2021

- ongoing, [see poster Lukas Vanpoucke](#)
 - relation to PDEs
 - least squares identification
 - open problem: minimal parameterization of mD difference equations
3. globally optimal LS identification of multiple-output systems
tools: functional analysis, operator theory, behavioral framework
- inner functions
 - Beurling-Lax theorem
 - isometric state space models
 - norm preserving behavioral models
4. globally optimal H_2 model reduction
- by solving an MEP⁶
proof-of-concept: order 2 \rightarrow order 1
 - improvements, ongoing, [see poster Sibren Lagauw](#)
based on Walsh's theorem: order 7 \rightarrow order 3

⁶Oscar M. Agudelo, Christof Vermeersch and Bart De Moor, Globally Optimal H_2 -Norm Model Reduction: A Numerical Linear Algebra Approach, IFAC PapersOnLine vol. 54, no. 9, pp. 564–571, July 2021