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Polynomial Eigenvalue Solutions to Minimal Problems in Computer Vision

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Minimal Problems

Minimal problems in computer vision arise when computing geometrical models from image data. They often lead to solving systems of algebraic equations.

P3P problem
5-pt relative pose problem
Hand-eye

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Minimal Problems

Since the solutions of a minimal problem are roots of a system of polynomial equations, minimal problems can be solved by the following two frequently used methods:

- 1. Grobner Bases
- 2. Polynomial Eigenvalue Solution

Contents

- Outline of the Polynomial Eigenvalue Solution of a system of polynomial equations
- Polynomial Eigenvalue Problems
- Transformation of Systems of Polynomial Equations to a PEP Macaulay's Resultant-Based Method
 - Resultant-Based Method proposed in this paper

Consider a system of equations

$$f_1(\mathbf{x}) = \dots = f_m(\mathbf{x}) = 0,$$

which is given by a set of m polynomials $F = \{f_1, \ldots, f_m | f_i \in \mathbb{C}[x_1, \ldots, x_n]\}$ in n variables $\mathbf{x} = (x_1, \ldots, x_n)$ over the field of complex numbers.

If the number of independent equations equals the number of unknowns, this system have a finite number of solutions. In this case, this system is called a zero-dimensional polynomial system.

Let

$$f_i(\mathbf{x}) = \mathbf{c}_i(x_1)\mathbf{v}(x_2,\dots,x_n)$$
 Hide x_1

Then, the original polynomial equations can be written as:

$$\begin{cases} f_1(\mathbf{x}) = \mathbf{c}_1(x_1)\mathbf{v}(x_2, \dots, x_n) \\ \vdots \\ f_m(\mathbf{x}) = \mathbf{c}_m(x_1)\mathbf{v}(x_2, \dots, x_n) \end{cases}$$

$$\mathbf{c}(x_1)\mathbf{v} = \left(egin{array}{c} \mathbf{c}_1(x_1) \\ \vdots \\ \mathbf{c}_m(x_1) \end{array}
ight) \mathbf{v} = \mathbf{0}$$

Polynomial eigenvalue solution

$$\det(\mathbf{c}(x_1)) = 0$$

The roots are called eigenvalues.

$$\mathbf{c}(x_1)\mathbf{v} = \left(egin{array}{c} \mathbf{c}_1(x_1) \\ \vdots \\ \mathbf{c}_m(x_1) \end{array}
ight) \mathbf{v} = \mathbf{0}$$

The null vectors are called eigenvectors. From the eigenvectors, solutions of other variables can be found

A simple case

$$\begin{cases} xy - 1 = 0 \\ x + 1 - y = 0 \end{cases}$$

$$\begin{pmatrix} x & 1 \\ 1 & x+1 \end{pmatrix} \begin{pmatrix} y \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\det \left(\begin{array}{cc} x & 1 \\ 1 & x+1 \end{array} \right) = 0$$

$$x(x+1) - 1 = 0$$

Eigenvalues:
$$x_1=rac{-1+\sqrt{5}}{2}$$
 $x_2=rac{-1-\sqrt{5}}{2}$

Hide x
$$\begin{pmatrix} x & 1 \\ 1 & x+1 \end{pmatrix} \begin{pmatrix} y \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$
 Eigenvectors:
$$\mathbf{v}_1 = \begin{pmatrix} -\frac{1+\sqrt{5}}{2} \\ 1 \end{pmatrix} y_1 = \frac{1+\sqrt{5}}{2}$$

$$\det \begin{pmatrix} x & 1 \\ 1 & x+1 \end{pmatrix} = 0$$

$$\mathbf{v}_2 = \begin{pmatrix} -\frac{1-\sqrt{5}}{2} \\ 1 \end{pmatrix} y_2 = \frac{1-\sqrt{5}}{2}$$

Polynomial eigenvalue problems are problems of the form

$$C(\lambda)\mathbf{v} = 0,\tag{1}$$

where \mathbf{v} is a vector of monomials in all variables except for λ and $C(\lambda)$ is a matrix polynomial in variable λ defined as

$$C(\lambda) \equiv \lambda^{l} C_{l} + \lambda^{l-1} C_{l-1} + \dots + \lambda C_{1} + C_{0}, \qquad (2)$$

with $n \times n$ coefficient matrices C_j [1].

[1] Z. Bai, J. Demmel, J. Dongorra, A. Ruhe, and H. van der Vorst, Templates for the Solution of Algebraic Eigenvalue Problems. SIAM, 2000.

Solution:

From a PEP

$$C(\lambda)\mathbf{v} = 0$$

to a Generalized Eigenvalue Problems(GEP)

$$A \mathbf{y} = \lambda B \mathbf{y}. \tag{3}$$

High order PEPs of degree l,

$$(\lambda^{l}\mathbf{C}_{l} + \lambda^{l-1}\mathbf{C}_{l-1} + \cdots + \lambda\mathbf{C}_{1} + \mathbf{C}_{0})\mathbf{v} = 0,$$

can be transformed to the generalized eigenvalue problem (3). Here,

$$\mathtt{A} = \left(egin{array}{cccccc} \mathtt{I} & \mathtt{0} & \mathtt{1} & \mathtt{0} & \mathtt{...} & \mathtt{0} \ \mathtt{0} & \mathtt{0} & \mathtt{I} & \mathtt{...} & \mathtt{0} \ \mathtt{...} & \mathtt{...} & \mathtt{...} & \mathtt{...} \ -\mathtt{C}_0 & -\mathtt{C}_1 & -\mathtt{C}_2 & \mathtt{...} & -\mathtt{C}_{l-1} \end{array}
ight), \ \mathtt{B} = \left(egin{array}{ccccc} \mathtt{I} & & & & & & & & & & & & & \\ & \mathtt{I} & & & & & & & & & & & & & \\ & & \mathtt{I} & & & & & & & & & & & \\ & & & \mathtt{C}_l \end{array}
ight), \quad \mathbf{y} = \left(egin{array}{c} \mathbf{v} \\ \lambda \mathbf{v} \\ \mathtt{...} \\ \lambda^{l-1} \mathbf{v} \end{array}
ight).$$

$$\begin{pmatrix} 0 & \mathbf{I} & 0 & \dots & 0 \\ 0 & 0 & \mathbf{I} & \dots & 0 \\ \dots & \dots & \dots & \dots \\ -\mathbf{C}_0 & -\mathbf{C}_1 & -\mathbf{C}_2 & \dots & -\mathbf{C}_{l-1} \end{pmatrix} \begin{pmatrix} \mathbf{v} \\ \lambda \mathbf{v} \\ \dots \\ \lambda^{l-1} \mathbf{v} \end{pmatrix} = \lambda \begin{pmatrix} \mathbf{I} & & \\ & \ddots & \\ & & \mathbf{I} \\ & & & \mathbf{C}_l \end{pmatrix} \begin{pmatrix} \mathbf{v} \\ \lambda \mathbf{v} \\ \dots \\ \lambda^{l-1} \mathbf{v} \end{pmatrix}$$
 If \mathbf{C}_l is nonsingular and well conditioned, multiply
$$\begin{pmatrix} \mathbf{I} & & \\ & \ddots & \\ & & \mathbf{I} \\ & & \mathbf{C}_l^{-1} \end{pmatrix}$$
 on both sides.

Often, ${f C}_l$ is singular but ${f C}_0$ is regular and well-conditioned. In this case, let $~eta=1/\lambda$.

$$oldsymbol{A} egin{pmatrix} \mathbf{v} \\ \lambda \mathbf{v} \\ \dots \\ \lambda^{l-1} \mathbf{v} \end{pmatrix} = \lambda egin{pmatrix} \mathbf{v} \\ \lambda \mathbf{v} \\ \dots \\ \lambda^{l-1} \mathbf{v} \end{pmatrix}$$

$$\mathtt{A} = egin{pmatrix} 0 & \mathtt{I} & 0 & \ldots & 0 \ 0 & 0 & \mathtt{I} & \ldots & 0 \ \ldots & \ldots & \ldots & \ldots & \ldots \ -\mathsf{C}_0^{-1}\mathsf{C}_l & -\mathsf{C}_0^{-1}\mathsf{C}_{l-1} & -\mathsf{C}_0^{-1}\mathsf{C}_{l-2} & \ldots & -\mathsf{C}_0^{-1}\mathsf{C}_1 \end{pmatrix}$$

For the roots of the polynomial equations, the above equation holds, but not for all PES's are solutions of the original problem.

Find the solutions either by :

- 1. testing all monomial dependencies in v, or
- 2. by substituting the solutions to the original equations and checking if they are satisfied.

^{*}Screen out the pseudo solutions of this relaxation method, not the pseudo solutions of the original problem.

Consider a system of equations

$$f_1(\mathbf{x}) = \dots = f_m(\mathbf{x}) = 0,$$

In some cases, for some x_j , let say x_1 , the above system can directly be rewritten to a polynomial eigenvalue problem:

$$C(x_1)\mathbf{v}=0,$$

where $C(x_1)$ is a matrix polynomial with square $m \times m$ coefficient matrices and \mathbf{v} is a vector of s monomials in variables x_2, \ldots, x_n , i.e., monomials of the form $\mathbf{x}^{\alpha} = x_2^{\alpha_2} x_3^{\alpha_3} \ldots x_n^{\alpha_n}$. In this case, the number of monomials s is equal to the number of equations m, i.e., s = m.

However, we are not always lucky. The number of monomials may be larger than that of the polynomials. New linearly independent polynomial equations are needed.

Macaulay's Resultant-Based Method

$$f_1(x_1,\ldots,x_n) = \cdots = f_n(x_1,\ldots,x_n) = 0.$$

$$f_1, \ldots, f_n \in (\mathbb{C}[x_1])[x_2, \ldots, x_n].$$

Let the degrees of these equations in variables x_2, \ldots, x_n be d_1, d_2, \ldots, d_n , respectively.

Homogenization:
$$F_i=x_{n+1}^{d_i}\ f_i(rac{x_2}{x_{n+1}},\ldots,rac{x_n}{x_{n+1}})$$
 .

Macaulay's Resultant-Based Method

Let
$$d = \sum_{i=1}^{n} (d_i - 1) + 1 = \sum_{i=1}^{n} d_i - n + 1.$$

take the set of all monomials $~\mathbf{x}^lpha=x_2^{lpha_2}x_3^{lpha_3}\dots x_n^{lpha_n}x_{n+1}^{lpha_{n+1}}$ $\sum_{i=2}^{n+1}lpha_i=d$

$$S_1 = \{\mathbf{x}^{lpha} : |lpha| = d, \ x_2^{d_1} |\mathbf{x}^{lpha}\},$$
 $S_2 = \{\mathbf{x}^{lpha} : |lpha| = d, \ x_2^{d_1} \not\mid \mathbf{x}^{lpha} \ ext{but} \ x_3^{d_2} \mid \mathbf{x}^{lpha}\},$
 \dots

$$S_n = \{ \mathbf{x}^{\alpha} : |\alpha| = d, \ x_2^{d_1}, \dots, x_n^{d_{n-1}} \not\mid ; \mathbf{x}^{\alpha} \text{ but } x_{n+1}^{d_n} \mid \mathbf{x}^{\alpha} \},$$

Macaulay's Resultant-Based Method

Generalize the original system

$$\mathbf{x}^{\alpha}/x_2^{d_1} \ F_1 = 0 \quad \text{for all } \mathbf{x}^{\alpha} \in S_1$$

$$\cdots$$

$$\mathbf{x}^{\alpha}/x_{n+1}^{d_n} \ F_n = 0 \quad \text{for all } \mathbf{x}^{\alpha} \in S_n.$$

Dehomogenization:

$$x_{n+1} = 1$$

Macaulay's Resultant-Based Method

Disadvantage: designed for dense and small problems.

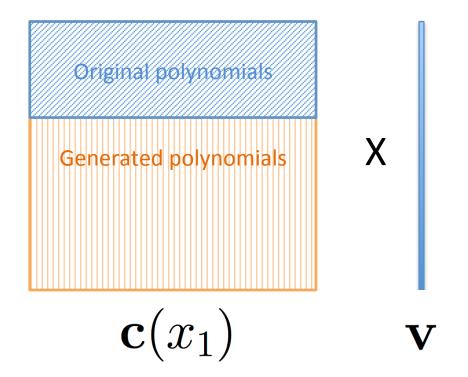
Resultant-Based Method proposed by Kukelova et al

$$S_1 = \{\mathbf{x}^{lpha} : |lpha| = d, \ x_2^{d_1} | \mathbf{x}^{lpha} \},$$
 $S_2 = \{\mathbf{x}^{lpha} : |lpha| = d, \ x_2^{d_1}
mid \mathbf{x}^{lpha} \ ext{ but } \ x_3^{d_2} \mid \mathbf{x}^{lpha} \},$
 \dots
 $S_n = \{\mathbf{x}^{lpha} : |lpha| = d, \ x_2^{d_1}, \dots, x_n^{d_{n-1}}
mid \mathbf{x}^{lpha} \ ext{ but } \ x_{n+1}^{d_n} \mid \mathbf{x}^{lpha} \},$

$$egin{aligned} \overline{S_1} &= \{\mathbf{x}^lpha : |lpha| = d, \ x_2^{d_1} \mid \mathbf{x}^lpha \}, \ \overline{S_2} &= \{\mathbf{x}^lpha : |lpha| = d, \ x_3^{d_2} \mid \mathbf{x}^lpha \}, \ & \cdots \ \overline{S_n} &= \{\mathbf{x}^lpha : |lpha| = d, \ x_{n+1}^{d_n} \mid \mathbf{x}^lpha \}. \end{aligned}$$

Reducing the Size of the Polynomial Eigenvalue Problem

Removing Unnecessary Polynomials

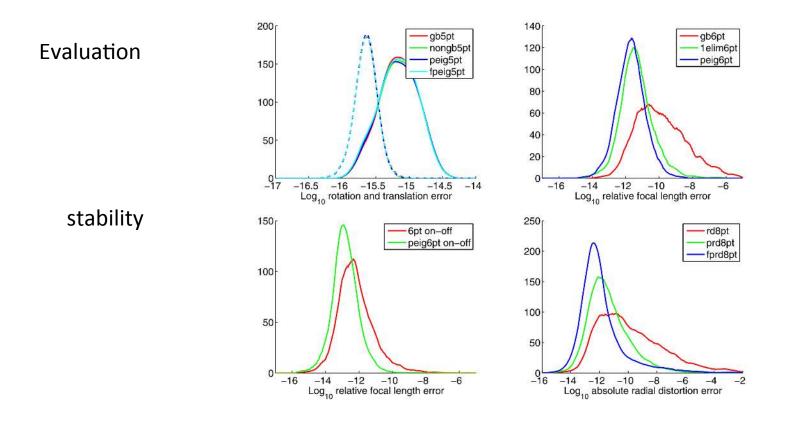


Reducing the Size of the Polynomial Eigenvalue Problem

Removing Zero Eigenvalues

$$\mathbf{A} = \begin{pmatrix} 0 & \mathbf{I} & 0 & \dots & 0 \\ 0 & 0 & \mathbf{I} & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ -\mathbf{C}_0^{-1}\mathbf{C}_l & -\mathbf{C}_0^{-1}\mathbf{C}_{l-1} & -\mathbf{C}_0^{-1}\mathbf{C}_{l-2} & \dots & -\mathbf{C}_0^{-1}\mathbf{C}_1 \end{pmatrix}$$

$$\mathbf{A} \begin{pmatrix} \mathbf{v} \\ \lambda \mathbf{v} \\ \dots \\ \lambda^{l-1} \mathbf{v} \end{pmatrix} = \lambda \begin{pmatrix} \mathbf{v} \\ \lambda \mathbf{v} \\ \dots \\ \lambda^{l-1} \mathbf{v} \end{pmatrix}$$



Time: ~us. Faster than Grobner Bases

The End, Thanks!