

P-solutions of linear and nonlinear interval parametric systems and applications

L. Kolev

Theoretical Electrical Eng., TU-Sofia, 1000 Sofia, Bulgaria,

E-mail: lkolev@tu-sofia.bg

Outline

- Introduction
- Properties of P-solutions
- Methods for determining P-solutions
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1. Introduction

Consider the following parametric systems of decreasing generality and complexity:

- nonlinear interval parametric (NLIP) system

$$f(x, p) = 0$$

$$f_i(x_1, \dots, x_n; p_1, \dots, p_m) = 0, \quad p_j \in \mathbf{p}_j = [-1, 1], \quad i = 1, \dots, n \quad (1.10)$$

x_j are the variables and p_j are the parameters

- linear interval parametric (LIP) system

$$f(x, p) = A(p)x - b(p) = 0 \quad , \quad p \in \mathbf{p} \quad (1.11)$$

- nonlinear parametric dependences (NLPD) systems

- linear parametric dependence (LPD) systems, where

$$a_{ij}(p) = \alpha_{ij} + \sum_{\mu=1}^m a_{ij\mu} p_{\mu}, \quad b_i(p) = \beta_i + \sum_{\mu=1}^m \beta_{i\mu} p_{\mu} \quad (1.11b)$$

We need the concept of the solution set Σ of (1.10) or (1.11) defined as follows

$$\Sigma = \{x : f(x, p) = 0, p \in P\} \quad (1.12)$$

Known “interval solutions” to (1.10) or (1.11): (i) interval hull (IH) solution \mathbf{x}^* : the smallest interval vector containing Σ ; (ii) outer interval (OI) solution \mathbf{x} : any interval vector enclosing \mathbf{x}^* , i.e. $\mathbf{x}^* \subseteq \mathbf{x}$ and (iii) inner estimation of the hull (IEH) solution ξ : an interval vector such that $\xi \subseteq \mathbf{x}^*$.

Recently, a new type of solution to LPD systems (1.11), (1.11b) has been introduced [Kolev 2014] which is of the following parametric form

$$\mathbf{x}(p) = Lp + \mathbf{a}, \quad p \in \mathcal{P} \quad (1.13)$$

where L is a real $n \times m$ matrix whereas \mathbf{a} is an interval vector. It is called a parameterized solution (p -solution) of (1.11). The p -solution is an outward linear approximation of Σ .

Iterative methods for determining $\mathbf{x}(p)$ were suggested for the LPD case in [Kolev 2014,2016 Rel. Comp.]; a simple direct method was proposed in [Kolev 2016 JACM]. Also the NLPD case was considered in [Skalna and Hladik 2017].

2. Properties of P-solutions

Let $\mathbf{x}(p)$ be a p -solution of system (1.10). Interval hull $\mathbf{x}(p)$ of $\mathbf{x}(p)$ - the smallest interval containing $\mathbf{x}(p)$.

- **Property P1.** The interval hull $\mathbf{x}(p)$ of $\mathbf{x}(p)$ is an outer interval solution \mathbf{x} of (1.10), i.e. $\mathbf{x} = \mathbf{x}(p)$.

Consider the i th component x_i^* of the IH solution \mathbf{x}^* to (1.10) and the i th component $x_i(p)$ of $\mathbf{x}(p)$

$$x_i(p) = \tilde{x}_i + L_i p + \hat{x}_i [-1, 1], \quad p \in \mathcal{P} \quad (3.1)$$

(\tilde{x}_i, \hat{x}_i - centre and radius, L_i - i th row of L). Compute the ends

$$\underline{x}_i = \tilde{x}_i - \sum_j |L_j| - \hat{x}_i, \quad \bar{x}_i = \tilde{x}_i + \sum_j |L_j| + \hat{x}_i \quad (3.2)$$

Introduce the intervals:

$$\mathbf{e}_i^{(l)} = [\underline{x}_i, \underline{x}_i + 2\hat{x}_i], \quad \mathbf{e}_i^{(u)} = [\bar{x}_i, \bar{x}_i - 2\hat{x}_i] \quad (3.4)$$

- **Property P2.** The intervals (3.4) contain the endpoints of \mathbf{x}_i^* , i.e.

$$\underline{x}_i^* \in \mathbf{e}_i^{(l)}, \quad \bar{x}_i^* \in \mathbf{e}_i^{(u)} \quad (3.5)$$

Thus, the intervals (3.4) provide two-sided bounds on the ends of \mathbf{x}_i^*

- **Property P3.** Introduce the intervals

$$\xi_i = \begin{cases} [e_i^{-(l)}, e_i^{(u)}], & \text{if } e_i^{-(l)} < e_i^{(u)} \\ \text{empty interval,} & \text{otherwise} \end{cases} \quad (3.6)$$

Then ξ_i determines the i th component of the IEH solution of (1.10). 7

3. Methods for determining a p -solution

All known methods refer to LIP systems and have polynomial time complexity. Here we only mention the

3.1. *Direct method of [Kolev 2016] (Method K)*

which requires roughly $N_k = n^4 m$ arithmetic operations

Comparison with method of [Skalna 2006] (Method S_k)

Methods K and S_k have the same complexity. However, unlike method K , method S_k has only Property P1.

3.2. *A new method*

It is applicable to the general type of NLIP systems

$$f(x, p) = 0, \quad y \in p \tag{3.32}$$

Computational scheme.

- 1). An outer interval solution \mathbf{x}^0 is computed (using some standard method).
- 2). The variables \mathbf{x}^0 and y in (3.32) are then treated as independent, so (3.32) is replaced by

$$f(\mathbf{x}, p) = 0, \quad \mathbf{x} \in \mathbf{x}^0, \quad p \in p \quad (3.34)$$

- 3). The variables x_i are written in normalized form

$$x_i = \check{x}_i + \hat{x}_i q_i, \quad q_i \in \mathbf{q}_i = [-1, 1] \quad (3.35a)$$

Using (3.35a), (3.34) is modified to

$$f(\mathbf{q}, p) = 0 \quad (3.37a)$$

4). Using affine arithmetic (AA) the nonlinear system (3.37a) is enclosed by the LIP system.

$$l(q, p) = \check{l} + A^q q + A^p p + \hat{l}[-1, 1] = 0, \quad q \in \mathbf{q}, \quad p \in \mathbf{p} \quad (3.37b)$$

where A^q and A^p are matrices of size $n \times n$ and $n \times m$

5). Solve for

$$q = \check{q} + Lp + \hat{q}[-1, 1], \quad p \in \mathbf{p} \quad (3.38)$$

6). Using (3.35a) and q we compute

$$\mathbf{x}(p) = \check{x} + Lp + \hat{x}[-1, 1], \quad p \in \mathbf{p} \quad (3.40)$$

7). Let \mathbf{x} be the interval hull of (3.40).

If
$$\mathbf{x} \subseteq \mathbf{x}^0 \quad (3.40a)$$

then $\mathbf{x}(p)$ obtained by (3.35a) to (3.40) is a parameterized solution of (3.32)

If \mathbf{x} is narrower than \mathbf{x}^0 and the reduction is larger than a threshold ε_r , then \mathbf{x} is renamed \mathbf{x}^0 and a new iteration can be resumed from (3.35) until some stopping criterion is met.

Numerical example with $n = 2$ and $m = 2$

The method was programmed in MATLAB environment using the toolbox IntLab.8 [Rump 1999]. The AA arithmetic was implemented by the *affari* toolbox [Rump and Kashiwagi 2015]

The system (3.34) is

$$f_1(x, y) = -y_1 x_1 + x_2^2 = 0, \quad (3.46a)$$

$$f_2(x, y) = y_2 f_{21}(x_1) + x_2^2 = 0, \quad f_{21}(x_1) = x_1(0.5x_1 - 1) - 1, \quad (3.46b)$$

$$y \in \mathbf{y}, \quad x \in \mathbf{x}^0 \quad (3.46c)$$

$$y_1 \in \mathbf{y}_1 = [15.8, 16.2], \quad y_2 \in \mathbf{y}_2 = [18, 19.6]$$

$$\mathbf{x}_1^0 = [1.2, 1.7], \quad \mathbf{x}_2^0 = [-5.8, -4.7]$$

Determining a p-solution

The related system (3.37b) is

$$\tilde{l} = \begin{pmatrix} -4.5137 \\ -1.7110 \end{pmatrix}, \quad A^q = \begin{pmatrix} -4.000 & -5.775 \\ 2.115 & -5.775 \end{pmatrix}, \quad A^p = \begin{pmatrix} 0.2900 & 0 \\ 0 & 1.1065 \end{pmatrix}, \quad \hat{l} = \begin{pmatrix} 0.20125 \\ 0.54750 \end{pmatrix}$$

Hence

$$\mathbf{x}(p) = \tilde{\mathbf{x}} + Lp + \hat{\mathbf{x}}[-1, 1], \quad p \in \mathbf{p} \quad (3.47)$$

$$\tilde{\mathbf{x}} = \begin{pmatrix} 1.5645850 \\ -4.9947248 \end{pmatrix}, \quad L = \begin{pmatrix} -0.0118561 & 0.0523712 \\ -0.0095526 & -0.0689328 \end{pmatrix}, \quad \hat{\mathbf{x}} = \begin{pmatrix} 0.0877044 \\ 0.1192227 \end{pmatrix} \quad (3.47a)$$

Finally

$$\mathbf{x} = \begin{pmatrix} [1.4768806, 1.6243233] \\ [-5.1139475, -4.8755021] \end{pmatrix} \quad (3.48)$$

and it is seen that $\mathbf{x} \subset \mathbf{x}^0$

Thus, the LIP form (3.47) determines the p -solution sought.

The subsequent iterations of method M1 yield better enclosures.

Convergence is reached in $k = 4$ iterations (relative reduction of $\hat{\mathbf{x}}_i^0$ to $\hat{\mathbf{x}}_i$ is less 1%)

Property 2. Determining the two-sided bound

We illustrate with $\mathbf{e}_2^l = [\underline{e}_2^l, \bar{e}_2^l]$. For $k=4$

$$\mathbf{e}_2^l = [-5.1139, -4.8755]$$

The lower end \underline{x}_2^* of the IH solution \mathbf{x}_2^* is approximately

$$\underline{x}_2^* = -5.0884$$

and it is seen that indeed

$$\underline{x}_2^* \in \mathbf{e}_2^l$$

4. A class of global optimization problems

Find the global minimum

$$f_0^* = \min f_0(x, p) \quad (4.1a)$$

$$f_i(x, p) = 0, \quad p \in \mathbf{p}, \quad i = 1, 2, \dots, n \quad (4.1b)$$

The p -solution of (4.1b) – effective to solve (4.1).

Each individual problem is set up by specifying the functions $f_0(x, p)$ and $f(x, p)$. For instance:

Linear programming (LP) problem

$$f_0(p) = c^T(p) x(p)$$

where the constraint is the LIP system $A(p)x = b(p)$ [Kolev 2016]

Numerical example

$$c^T = (1, 1, 1)$$

If $c = \pm e_k$ (e_k is the k -th column of the identity matrix) the LP problem determines the lower/upper end of x_k^*

A method for solving such problems is available

[Kolev and Skalna 2017] which is based on expressing

$$f_0(p) = \sum_i x_i(p), \quad p \in \mathcal{P}$$

Numerical evidence shows that it is superior to standard methods wrt:

- enclosure efficiency: tightness of the approximation of the solution
- computational cost;
- applicability radius [Kolev 2014, 2016]: largest radius of the box within which the respective method is still applicable

5. Applications

- *Worst-case tolerance analysis for electric circuits* [Kolev 2002]
 - *Direct current circuits*
 - *Alternating current circuits*

- *Power consumption analysis* [Kolev 2011, 2013]
- *Truss analysis* [Muhanna et. Al. 2004, 2005; Popova 2006]

Each of the above problems consists in determining or bounding a component x_k^* of the corresponding hull solution x^* . The latter problem can be solved as corresponding simple interval linear programming ($c = \pm e_k$) using the approach of Section 4.

- *Eigenvalue range determination*

The standard method [Kolev 2010] can be improved using corresponding p -solutions

6. Conclusions

- The new methods employing p -solutions are better than their standard counterparts.
- NLIP problems can now be addressed using the new method of Section 3.2.
- Further numerical evidence would be welcome.
- It is hoped that the present survey will help promote the use of the new p -solution approach for solving various engineering problems involving uncertainties and risks.

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