

Guaranteed Bounds for Solution of Parameter Dependent System of Equations

Andrew Pownuk¹, Iwona Skalna², and Jazmin Quezada¹

1 - The University of Texas at El Paso, El Paso, Texas, USA

2 - AGH University of Science and Technology, Krakow, Poland

22th Joint UTEP/NMSU Workshop on Mathematics,
Computer Science, and Computational Sciences

Outline

- 1 Solution Set
- 2 Interval Methods
- 3 Optimization methods
- 4 New Approach
- 5 Example 1
- 6 Example 2
- 7 Example 3
- 8 Conclusions

Parameter dependent Boundary Value Problem

$$A(p)u = f(p), u \in V(p), p \in P$$

Exact solution

$$\underline{u} = \inf_{p \in P} u(p), \bar{u} = \sup_{p \in P} u(p)$$

$$u(x, p) \in [\underline{u}(x), \bar{u}(x)]$$

Approximate solution

$$\underline{u}_h = \inf_{p \in P} u_h(p), \bar{u}_h = \sup_{p \in P} u_h(p)$$

$$u_h(x, p) \in [\underline{u}_h(x), \bar{u}_h(x)]$$

Mathematical Models in Engineering

Solution Set

Interval
Methods

Optimization
methods

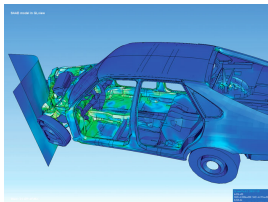
New Approach

Example 1

Example 2

Example 3

Conclusions



Linear and nonlinear equations.

Multiphysics (solid mechanics, fluid mechanics etc.)

Ordinary and partial differential equations, variational equations, variational inequalities, numerical methods, programming, visualizations, parallel computing etc.

Two point boundary value problem

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Sample problem

$$\begin{cases} -(g(x, p)u'(x))' = f(x, p) \\ u(0) = 0, u(1) = 0 \end{cases}$$

and $u_h(x)$ is finite element approximation given by a weak formulation

$$\int_0^1 g(x, p)u_h'(x)v'(x)dx = \int_0^1 f(x, p)v(x)dx, \forall v \in V_h^{(0)}$$

or

$$a(u_h, v) = l(v), \forall v \in V_h^{(0)} \subset H_0^1$$

where $u_h(x) = \sum_{i=1}^n u_i \varphi_i(x)$ and $\varphi_i(x_j) = \delta_{ij}$.

The Finite Element Method

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Approximate solution

$$\int_0^1 g(x, p) u'_h(x) v'(x) dx = \int_0^1 f(x, p) v(x) dx$$

$$\sum_{j=1}^n \left(\sum_{i=1}^n \int_0^1 g(x, p) \varphi_i(x) \varphi_j(x) dx u_i - \int_0^1 f(x, p) \varphi_j(x) dx \right) v_j = 0$$

Final system of equations (for one element) $Ku = q$ where

$$K_{i,j} = \int_0^1 g(x, p) \varphi_i(x) \varphi_j(x) dx, q_i = \int_0^1 f(x, p) \varphi_i(x) dx$$

Global Stiffness Matrix

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Global stiffness matrix

$$\sum_{p=1}^n \left(\sum_{q=1}^n \sum_{e=1}^{n_e} \sum_{i=1}^{n_u^e} \sum_{j=1}^{n_u^e} U_{j,p}^e \int_{\Omega_e} g(x, p) \frac{\partial \varphi_i^e(x)}{\partial x} \frac{\partial \varphi_j^e(x)}{\partial x} dx U_{i,q}^e u_q - \right.$$

$$\left. \sum_{q=1}^n \sum_{e=1}^{n_e} \sum_{i=1}^{n_u^e} \sum_{j=1}^{n_u^e} U_{j,p}^e \int_{\Omega_e} f(x, p) \varphi_i^e(x) \varphi_j^e(x) dx \right) v_p = 0$$

Final system of equations

$$K(p)u = q(p) \Rightarrow F(u, p) = 0$$

Solution Set

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Nonlinear equation $F(u, p) = 0$ for $p \in P$.

$$F : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n$$

Implicit function $u = u(p) \Leftrightarrow F(u, p) = 0$

$$u(P) = \{u : F(u, p) = 0, p \in P\}$$

Interval solution

$$\underline{u}_j = \min\{u : F(u, p) = 0, p \in P\}$$

$$\bar{u}_j = \max\{u : F(u, p) = 0, p \in P\}$$

Interval Methods

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

A. Neumaier, Interval Methods for Systems of Equations (Encyclopedia of Mathematics and its Applications, Cambridge University Press, 1991).

Z. Kulpa, A. Pownuk, and I. Skalna, Analysis of linear mechanical structures with uncertainties by means of interval methods, Computer Assisted Mechanics and Engineering Sciences, 5, 443-477, 1998.

V. Kreinovich, A.V.Lakeyev, and S.I. Noskov. Optimal solution of interval linear systems is intractable (NP-hard). Interval Computations, 1993, 1, 6-14.

Interval Methods

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

T. Burczynski, J. Skrzypczyk, Fuzzy aspects of the boundary element method, Engineering Analysis with Boundary Elements, Vol.19, No.3, pp. 209-216, 1997

A. Neumaier and A. Pownuk, Linear Systems with Large Uncertainties, with Applications to Truss Structures, Journal of Reliable Computing, 13(2), 149-172, 2007.

Muhanna, R. L. and R. L. Mullen. Uncertainty in Mechanics Problems Interval-Based Approach, Journal of Engineering Mechanics 127(6), 557-566, 2001.

I. Skalna, A method for outer interval solution of systems of linear equations depending linearly on interval parameters, Reliable Computing, 12, 2, 107-120, 2006.

Optimization methods

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Interval solution

$$\underline{u}_i = \min\{u(p) : p \in P\} = \min\{u : F(u, p) = 0, p \in P\}$$

$$\bar{u}_i = \max\{u(p) : p \in P\} = \max\{u : F(u, p) = 0, p \in P\}$$

$$\underline{u}_i = \begin{cases} \min u_i \\ F(u, p) = 0 \\ p \in P \end{cases}, \bar{u}_i = \begin{cases} \max u_i \\ F(u, p) = 0 \\ p \in P \end{cases}$$

KKT Conditions

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Nonlinear optimization problem for $f(x) = x_i$

$$\begin{cases} \min_x f(x) \\ h(x) = 0 \\ g(x) \geq 0 \end{cases}$$

Lagrange function $L(x, \lambda, \mu) = f(x) + \lambda^T h(x) - \mu^T g(x)$

Optimality conditions can be solved by the Newton method.

$$\begin{cases} \nabla_x L = 0 \\ \nabla_\lambda L = 0 \\ \mu_i \geq 0 \\ \mu_i g_i(x) = 0 \\ h(x) = 0 \\ g(x) \geq 0 \end{cases}$$

KKT Conditions - Newton Step

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

$$F'(X)\Delta X = -F(X)$$

$$F'(X) = \begin{bmatrix} (\nabla_x^2 f(x) + \nabla_x^2 h(x)y)_{n \times n} & \nabla_x h(x)_{n \times m} & -I_{n \times n} \\ (\nabla_x h(x))^T_{m \times n} & 0_{n \times m} & 0_{m \times n} \\ Z_{n \times n} & 0_{n \times m} & X_{m \times n} \end{bmatrix}$$

$$\Delta X = \begin{bmatrix} \Delta x \\ \Delta y \\ \Delta z \end{bmatrix}, X = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$$F(X) = - \begin{bmatrix} \nabla_x f(x) + \nabla_x h^T(x)y - z \\ h(x) \\ XYe - \mu_k e \end{bmatrix}$$

Steepest Descent Method

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

In order to find maximum/minimum of the function u it is possible to apply the steepest descent algorithm.

- 1 Given x_0 , set $k = 0$.
- 2 $d^k = -\nabla f(x_k)$. If $d^k = 0$ then stop.
- 3 Solve $\min_{\alpha} f(x_k + \alpha d^k)$ for the step size α_k . If we know second derivative H then $\alpha_k = \frac{d_k^T d_k}{d_k^T H(x_k) d_k}$.
- 4 Set $x_{k+1} = x_k + \alpha_k d_k$, update $k = k + 1$. Go to step 1.

I. Skalna and A. Pownuk, Global optimization method for computing interval hull solution for parametric linear systems, International Journal of Reliability and Safety, 3, 1/2/3, 235-245, 2009.

New Approach for Finding Guaranteed Bounds

Solution Set

Interval
Methods

Optimization
methods

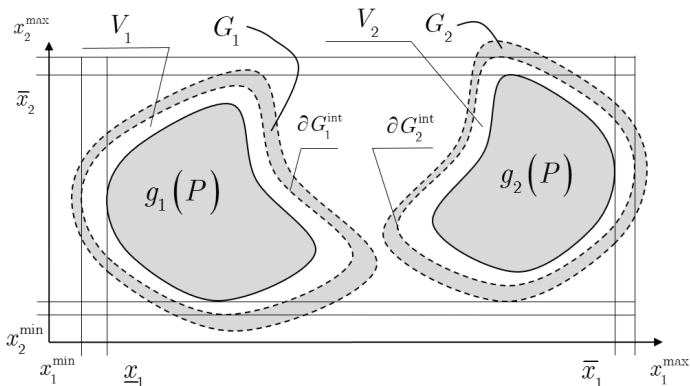
New Approach

Example 1

Example 2

Example 3

Conclusions



$$g_1(P) \subset V_1, g_2(P) \subset V_2$$
$$\partial V_1 = \partial G_1^{int}, \partial V_2 \subset \partial G_2^{int}$$

New Approach for Finding Guaranteed Bounds

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Theorem

Let's assume that $g : P \rightarrow \mathbb{R}$ is a continuous function, P is a path-connected, compact subset of \mathbb{R} , then

$g(P) = \{g(p) : p \in P\} = [g(p_{min}), g(p_{max})] = [x_{min}, x_{max}]$ is a closed interval and $p_{min}, p_{max} \in P$,

$x_{min} = \inf\{g(x) : p \in P\}, x_{max} = \sup\{g(x) : p \in P\}$.

New Approach for Finding Guaranteed Bounds

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Theorem

Let's assume that $g : P \rightarrow \mathbb{R}$ is a continuous function, P is a path-connected, compact subset of \mathbb{R}^m , we know at least one value $x_0 = g(p_0)$ such that $p_0 \in P$, and exists some $\varepsilon > 0$ such that $x_0 + \Delta x \notin g(P)$ for all $\Delta x \in (0, \varepsilon]$, then $x_0 = g(p_{max}) = g_{max}$ is a guaranteed upper bound of the set $g(P)$.

New Approach for Finding Guaranteed Bounds

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Theorem

Let's assume that $g : P \rightarrow \mathbb{R}^n$ is a continuous function that is defined as a unique solution of the equation $f(x, p) = 0$, P is a path-connected, compact subset of \mathbb{R}^m , we know at least one value $x_0 = g(p_0)$ such that $p_0 \in P$, and exists some $\varepsilon > 0$ such that $x_{0,i} + \Delta x_i \notin g_i(P)$ for all $\Delta x_i \in (0, \varepsilon]$, then $x_{0,i} = g_i(p_{max}) = g_{i,max}$ is a guaranteed upper bound of the set $g_i(P) = [g_{i,min}(P), g_{i,max}(P)]$.

If the equation $f(x, p) = 0$ has multiple solutions $x = g_i(p)$ ($i = 1, \dots, s$), then

$$g(P) = g_1(P) \cup g_2(P) \cup \dots \cup g_s(P)$$

Example 1

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Let's consider the equation nonlinear equation with uncertain parameter

$$x^2 - 4p^2 = 0 \text{ for } p \in [1, 2].$$

Presented equation has two solutions $x = g_1(p) = 2p$ and $x = g_2(p) = -2p$.

Non-guaranteed solutions are

$$[\underline{x}_1, \bar{x}_1] = g_1([1, 2]) = [2, 4] \text{ and}$$

$$[\underline{x}_2, \bar{x}_2] = g_2([1, 2]) = [-4, -2].$$

Example 1

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Let's check if the number $x_1 = 4 + \Delta x$ is a solution for $\Delta x > 0$.

$$0 \in f(4 + \Delta x, [1, 2])$$

$$0 \in (4 + \Delta x)^2 - 4[1, 2]^2$$

$$0 \in 16 + 8\Delta x + \Delta x^2 - 4[1, 4]$$

$$0 \in 16 + 8\Delta x + \Delta x^2 - [4, 16]$$

$$0 \in [8\Delta x + \Delta x^2, 12 + 8\Delta x + \Delta x^2]$$

Last condition is not satisfied then $x_1 = 4 + \Delta x$ is not a solution for any $\Delta x > 0$ then $\bar{x} = 4$.

Example 1

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Let's check if the number $x_1 = 2 - \Delta x$ is a solution for $\Delta x > 0$.

$$0 \in f(2 - \Delta x, [1, 2])$$

$$0 \in (2 - \Delta x)^2 - 4[1, 2]^2$$

$$0 \in 4 - 4\Delta x + \Delta x^2 - 4[1, 2]$$

$$0 \in 4 - 4\Delta x + \Delta x^2 + [-8, -4]$$

$$0 \in [-4 - 4\Delta x + \Delta x^2, -4\Delta x + \Delta x^2]$$

For small Δx it is possible to neglect the quadratic term and $-4\Delta x + \Delta x^2 < 0$. Last condition is not satisfied then $x_1 = 2 - \Delta x$ is not a solution for small $\Delta x > 0$ then $\underline{x} = 2$. The interval solution $[\underline{x}_1, \bar{x}_1] = [2, 4]$ is guaranteed.

Example 2

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

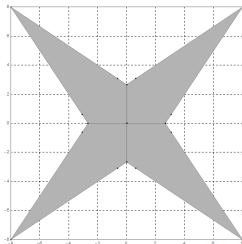
Example 3

Conclusions

Let's consider the following system of linear interval equations

$$\begin{cases} [-4, -3] x_1 + [-2, 2] x_2 = [-8, 8] \\ [-2, 2] x_1 + [-4, -3] x_2 = [-8, 8] \end{cases} \quad (1)$$

Let's assume that the non-guaranteed solution is $x_1 \in [-8, 8], x_2 \in [-8, 8]$.



Example 2

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Let $x_1 = 8 + \Delta x_1$ and $\Delta x_1 > 0$.

$$\begin{cases} a_{11}(8 + \Delta x_1) + a_{12}x_2 = b_1 \\ a_{21}(8 + \Delta x_1) + a_{22}x_2 = b_2 \end{cases}$$

$$\begin{cases} a_{11}(8 + \Delta x_1) + a_{12} \frac{b_2 - a_{21}(8 + \Delta x_1)}{a_{22}} = b_1 \\ x_2 = \frac{b_2 - a_{21}(8 + \Delta x_1)}{a_{22}} \end{cases}$$

$$a_{11}(8 + \Delta x_1) + \frac{a_{12}b_2}{a_{22}} - \frac{a_{12}a_{21}(8 + \Delta x_1)}{a_{22}} = b_1$$

$$(8 + \Delta x_1) \left(a_{11} - \frac{a_{12}a_{21}}{a_{22}} \right) + \frac{a_{12}b_2}{a_{22}} - b_1 = 0$$

Example 2

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

$$(8 + \Delta x_1) \left(a_{11} - \frac{a_{12}a_{21}}{a_{22}} \right) + \frac{a_{12}b_2}{a_{22}} - b_1 = 0$$

$$0 \in (8 + \Delta x_1) \left(a_{11} - \frac{a_{12}a_{21}}{a_{22}} \right) + \frac{a_{12}b_2}{a_{22}} - b_1$$

$$0 \in (8 + \Delta x_1) \left[-\frac{16}{3}, -\frac{5}{3} \right] + \left[-\frac{16}{3}, \frac{16}{3} \right] - [-8, 8]$$

$$0 \in \left[-\frac{128}{3} - \frac{16}{3}\Delta x_1, -\frac{40}{3} - \frac{5}{3}\Delta x_1 \right] + \left[-\frac{40}{3}, \frac{40}{3} \right]$$

$$0 \in \left[-56 - \frac{16}{3}\Delta x_1, -\frac{5}{3}\Delta x_1 \right]$$

Then $x_1 = 8 + \Delta x_1$ is not a solution for $\Delta x_1 > 0$ and $x_1 = 8$.

Parametric Linear System of Equations

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Sample boundary value problem

$$\frac{d}{dx} \left(EA \frac{du}{dx} \right) = 0, u(0) = 0, EA \frac{du(L)}{dx} = P$$

After discretization

$$\begin{bmatrix} k_1 + k_2 & -k_2 & 0 & \dots & 0 & 0 \\ -k_2 & k_2 + k_3 & -k_3 & \dots & 0 & 0 \\ & & & \dots & & \\ 0 & 0 & 0 & \dots & k_{n-1} + k_n & -k_n \\ 0 & 0 & 0 & \dots & -k_n & k_n \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ \dots \\ u_{n-1} \\ u_n \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \dots \\ 0 \\ 0 \end{bmatrix}$$

Let $n = 2$ and $k_1, k_2 \in [\frac{1}{3}, \frac{1}{2}]$ and $P = 1$ then the non-guaranteed solution is $u_1 \in [2, 3], u_2 \in [4, 6]$.

Example 3

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Let's assume that $u_1 = 3 + \Delta u_1$, $\Delta u_1 > 0$ then

$$\begin{cases} (k_1 + k_2)(3 + \Delta u_1) - k_2 u_2 = 0 \\ -k_2(3 + \Delta u_1) + k_2 u_2 = P \end{cases}$$

$$\begin{cases} (k_1 + k_2)(3 + \Delta u_1) - k_2 u_2 = 0 \\ u_2 = \frac{P + k_2(3 + \Delta u_1)}{k_2} \end{cases}$$

$$(k_1 + k_2)(3 + \Delta u_1) - k_2 \frac{P + k_2(3 + \Delta u_1)}{k_2} = 0$$

$$(k_1 + k_2)(3 + \Delta u_1) - P - k_2(3 + \Delta u_1) = 0$$

$$k_1(3 + \Delta u_1) - P = 0$$

Example 3

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

By assumption $k_1 \in \left[\frac{1}{3}, \frac{1}{2}\right]$, $P = 1$ then

$$0 \in \left[\frac{1}{3}, \frac{1}{2}\right] (3 + \Delta u_1) - P,$$

$$0 \in \left[\frac{1}{3}(3 + \Delta u_1), \frac{1}{2}(3 + \Delta u_1)\right] - 1,$$

$$0 \in \left[\frac{1}{3}\Delta u_1, \frac{1}{2} + \frac{1}{2}\Delta u_1\right].$$

The last condition cannot be satisfied because $\Delta u_1 > 0$ consequently $u_1 = 3 + \Delta u_1$ cannot be a solution of the system for any $\Delta u_1 > 0$ and $3 = \bar{u}_1$ is guaranteed upper-bound of the solution u_1 .

Conclusions

Solution Set

Interval
Methods

Optimization
methods

New Approach

Example 1

Example 2

Example 3

Conclusions

Methodology presented in this paper can be applied for wide range of parameter dependent system of equations and eigenvalue problems.

The method can be applied not only for the solution of the equations with set-valued parameters but also for finding values of the functions that depends of such solutions which is very important in the practical applications.

By using theory from the presentation, in some cases, in order to use guaranteed bounds of the solution it is possible existing, well established computational methods and at the end prove that the solution is guaranteed.

Methodology presented in this presentation can be applied to selected solutions or to all solutions of the systems of nonlinear equations.