Wave polynomials in inverse problems

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Schedule

- * Introduction
- * Wave polynomials
- * Solving problems by means of Trefftz functions
- * Thermoelasticity problems
- * Conclusions

History

- **1926** Erich Trefftz, *Ein Gegenstük zum Ritz'schen Verfahren*. Proceedings 2nd International Congres of Applied Mechanics, Zurich, pp.131–137.
- **1978** Herrera I., Sabina F., *Connectivity as an alternative to boundary integral equations: Construction of bases*, Appl. Math. Phys. Sc., No. 75/5, 2059–2063.
- Other authors: Kupradze, Jirousek, Leon, Zienkiewicz, Zieliński, Kołodziej, Qin, Li, Lu, Hu, Cheng.

Continuous time:

- **1956** P.C.Rosenbloom, D.V. Widder 1D Heat polynomials,
- **1998–2000** M.J.Ciałkowski, K.Grysa, S.Hożejowska, L.Hożejowski Direct and inverse problems for heat conduction (1D–3D, several coordinates systems),
- 2000 M.J.Ciałkowski, A.Frąckowiak Trefftz functions for Laplace'a, Poisson'a, Helmholtza and 1D wave equations, Trefftz functions as FEM base functions,
- 2002–2005 A.Maciąg, J.Wauer 2D and 3D wave polynomials,
- 2006— A.Maciąg Wave polynomials in thermoelasticity problems, T–functions for beam and plate vibrations problems, several variants of FEMT.

Main idea

The key idea of the method is to determine functions satisfying a given differential equation (T-functions) and to fit the linear combination of them to the governing initial and boundary conditions (usually The Least Square Method).

Advantages of the method

- Approximate solution (a linear combination of the solving functions) satisfies the equation identically and depends continuously on all space variables and time.
- The method is flexible in terms of given boundary and initial conditions (discrete, missing).
- The usage of T-functions as base functions in FEM allows to use big time-space elements.

Wave polynomials

Let us consider a wave equation in $\tilde{\Omega} = \Omega \times (0, t)$, where $\Omega \subset \mathbb{R}^n$ and t denotes time:

$$\nabla^2 u = \frac{1}{v^2} \frac{\partial^2 u}{\partial t^2}.$$

For two-dimensional wave equation we take $P_{000} = 1, Q_{000} = 0$. Then the recurrent formulas for wave polynomials are:

$$P_{(n-k)k0} = \frac{1}{n} (-xQ_{(n-k-1)k0} - yQ_{(n-k)(k-1)0} - vtQ_{(n-k-2)k1} - vtQ_{(n-k)(k-2)1}),$$

$$P_{(n-k-1)k1} = \frac{1}{n} (-xQ_{(n-k-2)k1} - yQ_{(n-k-1)(k-1)1} - vtQ_{(n-k-1)k0}),$$

$$Q_{(n-k)k0} = \frac{1}{n} (xP_{(n-k-1)k0} + yP_{(n-k)(k-1)0} + vtP_{(n-k-2)k1} + vtP_{(n-k)(k-2)1}),$$

$$Q_{(n-k-1)k1} = \frac{1}{n} (xP_{(n-k-2)k1} + yP_{(n-k-1)(k-1)1} + vtP_{(n-k-1)k0}).$$

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$$P_{000} = 1, \quad Q_{100} = x, \quad Q_{010} = y, \quad Q_{001} = t, P_{200} = -\frac{x^2}{2} - \frac{t^2}{2},$$
$$P_{110} = -xy, \quad P_{101} = -xt, \quad P_{011} = -yt, P_{020} = -\frac{y^2}{2} - \frac{t^2}{2}, \dots,$$

For three-dimensional wave equation:

$$P_{(n-k-l)kl0} = -\frac{1}{n} (xQ_{(n-k-l-1)kl0} + yQ_{(n-k-l)(k-1)l0} + zQ_{(n-k-l)k(l-1)0} + vtQ_{(n-k-l-2)kl1} + vtQ_{(n-k-l)(k-2)l1} + vtQ_{(n-k-l)k(l-2)1}),$$

$$P_{(n-k-l-1)kl1} = -\frac{1}{n} (xQ_{(n-k-l-2)kl1} + yQ_{(n-k-l-1)(k-1)l1} + zQ_{(n-k-l-1)k(l-1)1} + vtQ_{(n-k-l-1)kl0}),$$

$$Q_{(n-k-l)kl0} = \frac{1}{n} (xP_{(n-k-l-1)kl0} + yP_{(n-k-l)(k-1)l0} + zP_{(n-k-l)k(l-1)0} + vtP_{(n-k-l-2)kl1} + vtP_{(n-k-l)(k-2)l1} + vtP_{(n-k-l)k(l-2)1}),$$

$$Q_{(n-k-l-1)kl1} = \frac{1}{n} (xP_{(n-k-l-2)kl1} + yP_{(n-k-l-1)(k-1)l1} + zP_{(n-k-l-1)k(l-1)1} + vtP_{(n-k-l-1)kl0}).$$

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e.g.

T-functions method

As an approximation of the wave equation we take:

$$u \approx w = \sum_{n=1}^{N} c_n V_n.$$

Because all polynomials V_n satisfy the wave equation, the linear combination w satisfies this equation as well. The coefficients c_n are chosen so that the error of fulfilling given boundary and initial conditions corresponding to the wave equation is minimized.

The method is convergent – more polynomials leads to better approximation.

Membrane's vibrations

Consider the free vibrations of a square membrane described by equation

$$\frac{\partial^2 u}{\partial t^2} = \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} \quad (x, y) \in (0, 1) \times (0, 1), \quad t > 0,$$

and

initial conditions: $u(x, y, 0) = u_0(x, y) = \frac{1}{20} \sin(x) \sin(\pi y), \quad \frac{\partial u(x, y, 0)}{\partial t} = 0,$

boundary conditions: u(0, y, t) = u(x, 0, t) = u(x, 1, t) = 0.

For simulation we take a condition on the boundary x = 1:

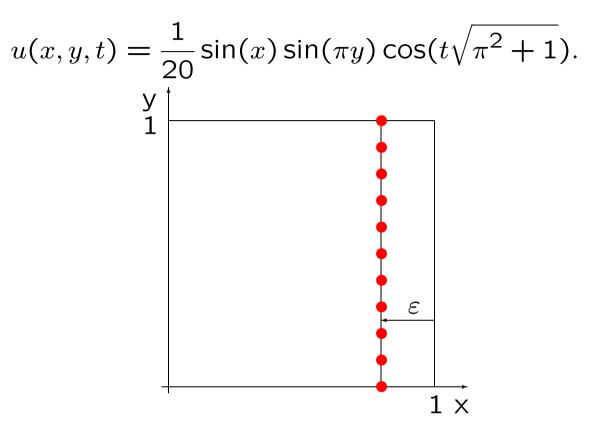
$$u(1, y, t) = \frac{1}{20} \sin(1) \sin(\pi y) \cos(t \sqrt{\pi^2 + 1}).$$

We assume that this condition is not known, but we know the internal responses of a membrane's deflection in distance ε from border x = 1:

$$u(1-\varepsilon, 0.1 \cdot (k-1), t) = u_k(t), \ k = 1, \dots, 11.$$

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Values of u_k for $k = 1 \dots 11$ were simulated from exact solution:



If $\varepsilon > 0$ we have an inverse problem and we search for a solution in the whole domain but especially u(1, y, t).

We check the quality of the identification of the boundary condition by calculating the average relative error in the whole time interval $(0, \Delta t)$, which is defined as:

$$E = \sqrt{\frac{\int_0^{\Delta t} [u(1,0.5,t) - w(1,0.5,t)]^2 dt}{\int_0^{\Delta t} [u(1,0.5,t)]^2 dt}} \cdot 100\%.$$

The values of error E in dependence on the distance ε and order N are shown in the table:

$N \backslash \varepsilon$	0	0.05	0.1	0.2	0.3	0.4	0.6
3	29.25	28.65	29.00	32.29	39.76	55.18	100.96
5	4.28	4.42	5.00	7.38	12.75	19.79	25.12
7	0.33	0.36	0.38	0.97	2.27	2.50	6.00

Disturbance and smoothing of data

Now we assume that the internal responses are given in discrete time points: $u_{kl} = u(1 - \varepsilon, (k - 1) \cdot 0.1, (l - 1) \cdot 0.1)$ where k, l = 1, ..., 11. This data are disturbed according to the formula:

$$u_{kl}^d = u_{kl}(1 + \xi_{kl}),$$

where ξ_{kl} are random numbers of normal distribution (N(0, 0.01)). In this case the relative error E has a value from circa 2000% for $\varepsilon = 0$ (direct problem) to circa 180000% for $\varepsilon = 0.8$.

As an approximation of disturbed internal responses we take:

$$u_{\varepsilon}(1-\varepsilon,y,t) = \sum_{n=1}^{N} c_n V_n(1-\varepsilon,y,t),$$

 V_n – wave polynomials. To chose coefficients c_n we minimize:

$$\sum_{k,l=1}^{11} [u_{\varepsilon}(1-\varepsilon,(k-1)\cdot 0.1,(l-1)\cdot 0.1)-u_{kl}^d]^2.$$

Then the values of the relative error E for the smoothed data are shown in the table:

ε	0	0.05	0.1	0.2	0.3	0.4	0.6
E	1.0	0.54	1.26	1.04	1.26	3.72	23.36

$N \backslash \varepsilon$	0	0.05	0.1	0.2	0.3	0.4	0.6
3	29.25	28.65	29.00	32.29	39.76	55.18	100.96
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7	0.33	0.36	0.38	0.97	2.27	2.50	6.00

If we take into account additional information concerning the velocity of the membrane in the same points (disturbed and smoothed by wave polynomials) we get:

	0						
E	0.32	0.31	0.32	0.42	0.5	0.6	3.48

Thermoelasticity problem

Let us consider thermoelasticity equations:

 $\mu \nabla^2 \mathbf{u} + (\lambda + \mu)$ grad div $\mathbf{u} = \rho \ddot{\mathbf{u}} + \gamma$ grad T

where **u** – displacement vector, ∇ – nabla operator, μ , λ – Lame constants, ρ – mass density, $\gamma = \frac{E}{1-2\nu}\alpha$, E – Young's modulus, ν – Poison's ratio, α – coefficient of thermal expansion.

The temperature field is described by the equation:

$$\frac{1}{\kappa}\frac{\partial T}{\partial t} = \nabla^2 T,$$

where κ – coefficient of thermal diffusivity. Equations are completed by initial and boundary conditions or "internal responses". The relation between displacements and stresses is described by:

$$\sigma_{ij} = 2\mu\varepsilon_{ij} + \lambda\varepsilon_{kk}\delta_{ij} - \gamma T\delta_{ij},$$

where $\varepsilon_{ij} = \frac{1}{2} \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)$ – strain tensor, δ_{ij} – Kronecker delta.

The system of thermoelasticity equations can be simplified by substitution: $\mathbf{u} = \operatorname{grad} \phi + \operatorname{rot} \Psi$ then we obtain:

$$(\nabla^2 - \frac{1}{v_1^2} \frac{\partial^2}{\partial t^2})\phi = mT, \quad (\nabla^2 - \frac{1}{v_2^2} \frac{\partial^2}{\partial t^2})\psi_i = 0, \qquad i = 1, 2, 3$$

where $v_1^2 = \frac{\lambda + 2\mu}{\rho}, \ v_2^2 = \frac{\mu}{\rho}, \ m = \frac{\gamma}{c_1^2 \rho}.$

We approximate the solution of the first equations by:

$$\phi \approx \widehat{\phi} = \sum_{n=1}^{N} c_n^0 V_n^0 + \phi_p.$$

For the second equation we take:

$$\psi_i \approx \widehat{\psi_i} = \sum_{n=1}^N c_n^i V_n^i, \quad i = 1, 2, 3$$

Here $V_n^i, i = 0, ..., 3$ are wave polynomials satisfying corresponding wave equation and ϕ_p is the particular solution. Coefficients c_n^i are determined by given conditions.

Particular solution

The particular solution ϕ_p for $L\phi = mT$ is calculated as $L^{-1}(mT)$, where $L = \nabla^2 - \frac{1}{v_1^2} \frac{\partial^2}{\partial t^2}$.

The temperature distribution can be approximated by linear combination of heat polynomials (or Taylor series). Therefore, we have to know how to calculate the inverse operator L^{-1} for monomials. It is easy to prove that for two space variables we have three forms of L^{-1} :

$$L_1^{-1}(x^k y^l t^m) = \frac{1}{(k+2)(k+1)} (x^{k+2} y^l t^m - l(l-1) Z_{(k+2)(l-2)m} + \frac{m(m-1)}{v_1^2} Z_{(k+2)l(m-2)}),$$

$$L_2^{-1}(x^k y^l t^m) = \frac{1}{(l+2)(l+1)} (x^k y^{l+2} t^m - k(k-1) Z_{(k-2)(l+2)m} + \frac{m(m-1)}{v_1^2} Z_{k(l+2)(m-2)}),$$

$$L_3^{-1}(x^k y^l t^m) = \frac{v_1^2}{(m+2)(m+1)} (-x^k y^l t^{m+2} + k(k-1)Z_{(k-2)l(m+2)} + l(l-1)Z_{k(l-2)(m+2)}).$$

For three space variables we have four forms of L^{-1} :

$$Z_{1}(x^{n}y^{k}z^{l}t^{m}) = \frac{1}{(n+2)(n+1)}(x^{n+2}y^{k}z^{l}t^{m} + \frac{m(m-1)}{v_{1}^{2}}Z_{(n+2)kl(m-2)} - k(k-1)Z_{(n+2)(k-2)lm} - l(l-1)Z_{(n+2)k(l-2)m}),$$

$$Z_{2}(x^{n}y^{k}z^{l}t^{m}) = \frac{1}{(k+2)(k+1)}(x^{n}y^{k+2}z^{l}t^{m} + \frac{m(m-1)}{v_{1}^{2}}Z_{n(k+2)l(m-2)} - n(n-1)Z_{(n-2)(k+2)lm} - l(l-1)Z_{n(k+2)(l-2)m}),$$

$$Z_{3}(x^{n}y^{k}z^{l}t^{m}) = \frac{1}{(l+2)(l+1)}(x^{n}y^{k}z^{l+2}t^{m} + \frac{m(m-1)}{v_{1}^{2}}Z_{nk(l+2)(m-2)} - n(n-1)Z_{(n-2)k(l+2)m} - k(k-1)Z_{n(k-2)(l+2)m}),$$

$$Z_{4}(x^{n}y^{k}z^{l}t^{m}) = \frac{v_{1}^{2}}{(m+2)(m+1)}(-x^{n}y^{k}z^{l}t^{m+2} + n(n-1)Z_{(n-2)kl(m+2)} + k(k-1)Z_{n(k-2)l(m+2)} + l(l-1)Z_{nk(l-2)(m+2)}].$$

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Test example

Consider the plane state of strain for $(x, y) \in (-1, 1) \times (-1, 1)$ when the strain tensor $\varepsilon_{ij} = \varepsilon_{ij}(x, y, t), (i, j = 1, 2)$ and $\varepsilon_{i3} = 0, (i = 1, 2, 3)$. Then the system of wave equations has a form:

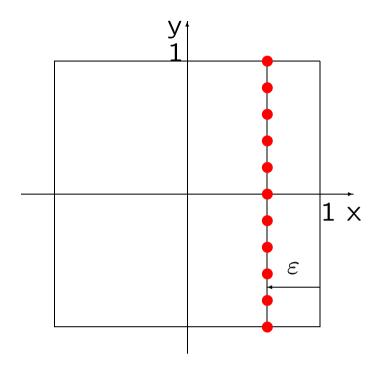
$$(\nabla^2 - \frac{1}{v_1^2} \frac{\partial^2}{\partial t^2})\phi(x, y, t) = mT(x, y, t), \quad (\nabla^2 - \frac{1}{v_2^2} \frac{\partial^2}{\partial t^2})\psi(x, y, t) = 0.$$

Displacements and stresses are given as:

$$\mathbf{u} = [u_x(x, y, t), u_y(x, y, t)] = \\ = \left[\frac{\partial \phi(x, y, t)}{\partial x} + \frac{\partial \psi(x, y, t)}{\partial y}, \frac{\partial \phi(x, y, t)}{\partial y} - \frac{\partial \psi(x, y, t)}{\partial x} \right], \\ \sigma_{xx} = (2\mu + \lambda) \frac{\partial u_x}{\partial x} + \lambda \frac{\partial u_y}{\partial y} - \gamma T, \\ \sigma_{xy} = \mu \left(\frac{\partial u_x}{\partial y} + \frac{\partial u_y}{\partial x} \right), \\ \sigma_{yy} = \lambda \frac{\partial u_x}{\partial x} + (2\mu + \lambda) \frac{\partial u_y}{\partial y} - \gamma T, \\ \sigma_{zz} = \lambda \left(\frac{\partial u_x}{\partial x} + \frac{\partial u_y}{\partial y} \right) - \gamma T.$$

Temperature and conditions: $T(x, y, t) = \frac{x^2}{2} + \frac{y^2}{2} + 2t$, $u_x(x,y,0) = \frac{mxy^2}{2}, \quad u_y(x,y,0) = \frac{mx^2y}{2},$ $\dot{u}_x(x,y,0) = \frac{2mx}{3} + \frac{\sqrt{2c_2}}{50000}\sin(x)\cos(y), \ \dot{u}_y(x,y,0) = \frac{2my}{3} - \frac{\sqrt{2c_2}}{50000}\cos(x)\sin(y),$ $u_x(-1, y, t) = -\frac{m}{2}(2t + y^2 - \frac{c_1^2 t^2}{2}) \pm \frac{\sin(1)}{50000}\cos(y)\sin(\sqrt{2}c_2 t),$ $u_x(x,-1,t) = u_x(x,1,t) = \frac{m}{3}(2xt + x - \frac{c_1^2xt^2}{2}) + \frac{\cos(1)}{50000}\sin(x)\sin(\sqrt{2}c_2t),$ $u_y(-1, y, t) = \frac{m}{3}(2yt + y - \frac{c_1^2 y t^2}{2}) - \frac{\cos(1)}{50000}\sin(y)\sin(\sqrt{2}c_2 t),$ $u_y(x,\pm 1,t) = \pm \frac{m}{2}(2t+x^2-\frac{c_1^2t^2}{2}) \mp \frac{\sin(1)}{50000}\cos(x)\sin(\sqrt{2}c_2t).$

The continuous conditions $u_x(1, y, t)$ and $u_y(1, y, t)$ are not known. Instead of that we know the values of them in discrete points $(x - \varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})$, k, l = 0, ..., 10 (internal responses).



Internal responses are simulated from the exact solution:

$$u_x = \frac{m}{3}(2xt + xy^2 - \frac{c_1^2xt^2}{2}) + \frac{1}{50000}\sin(x)\cos(y)\sin(\sqrt{2}c_2t),$$
$$u_y = \frac{m}{3}(x^2y + 2yt - \frac{c_1^2yt^2}{2}) - \frac{1}{50000}\cos(x)\sin(y)\sin(\sqrt{2}c_2t).$$

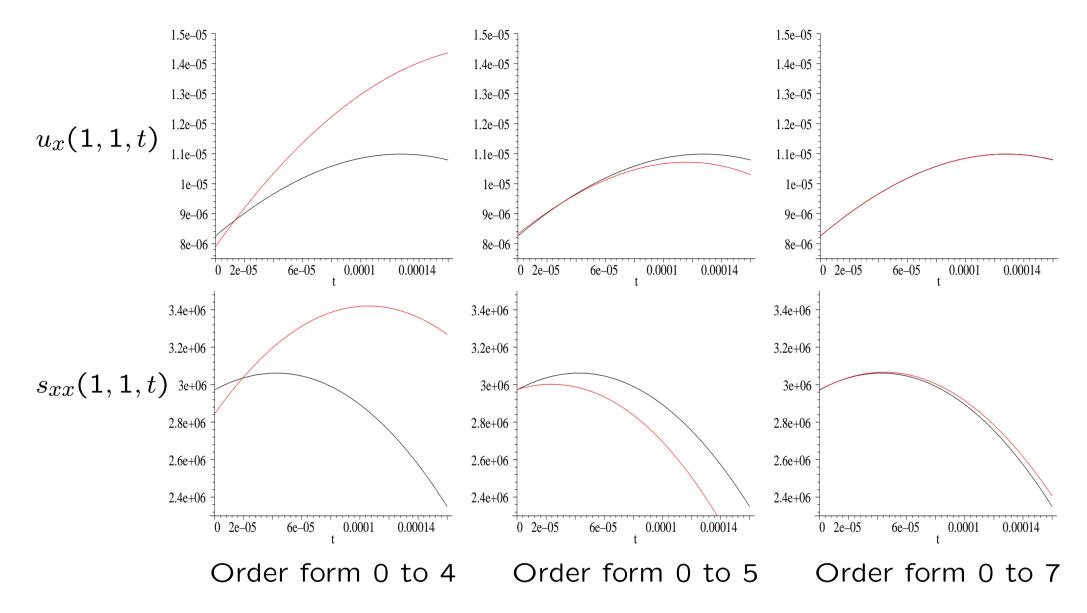
Approximate solution has a form: $\mathbf{u} \approx \widehat{\mathbf{u}} = \operatorname{grad} \widehat{\phi} + \operatorname{rot} \widehat{\Psi}$, where

$$\psi \approx \widehat{\psi} = \sum_{n=1}^{N} c_n^1 V_n^1, \ \phi \approx \widehat{\phi} = \sum_{n=1}^{N} c_n^0 V_n^0 + \phi_p.$$

The coefficients c_n^i are chosen so that the error of fulfilling given boundary and initial conditions is minimized:

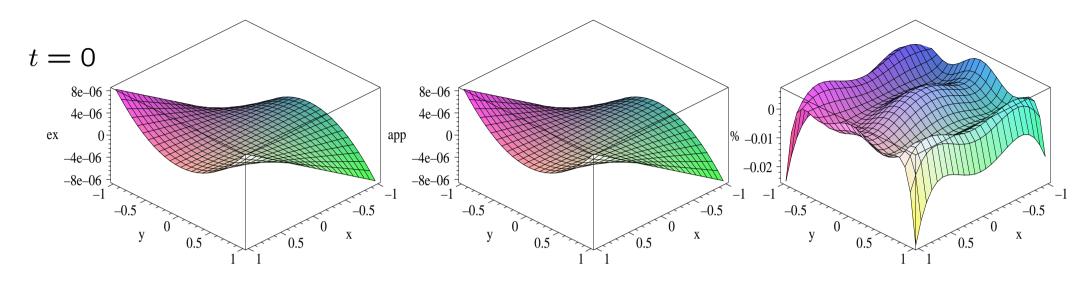
$$\begin{split} I &= \int_{-1}^{1} \int_{-1}^{1} \{ \underbrace{[\hat{u}_{x}(x,y,0) - u_{x}(x,y,0)]^{2} + [\hat{u}_{y}(x,y,0) - u_{y}(x,y,0)]^{2} + ...\} dy dx \\ &= \underbrace{\int_{-1}^{1} \int_{0}^{\Delta t} \{ [\hat{u}_{x}(-1,y,t) - u_{x}(-1,y,t)]^{2} + [\hat{u}_{y}(-1,y,t) - u_{y}(-1,y,t)]^{2} \} dt dy + ... \\ &= \underbrace{\int_{-1}^{1} \int_{0}^{\Delta t} \{ [\hat{u}_{x}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{x}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{1} \int_{0}^{10} \{ \underbrace{[\hat{u}_{x}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{x}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{1} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} } \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} } \\ &= \underbrace{\int_{-1}^{10} \int_{0}^{10} \int_{0}^{10} \{ \underbrace{[\hat{u}_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10}) - u_{y}(1-\varepsilon, -1 + \frac{k}{5}, \frac{l\Delta t}{10})]^{2} \} } \\ &= \underbrace{\int_{0}^{10} \int_{0}^{10} \int_{0}$$

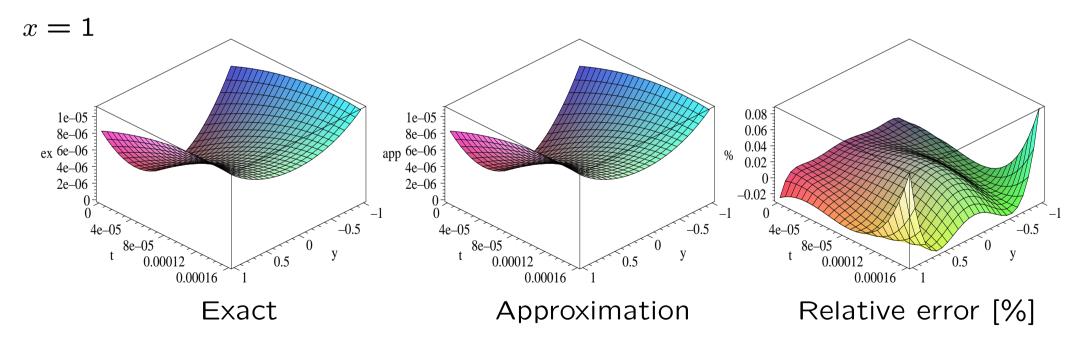
Convergence



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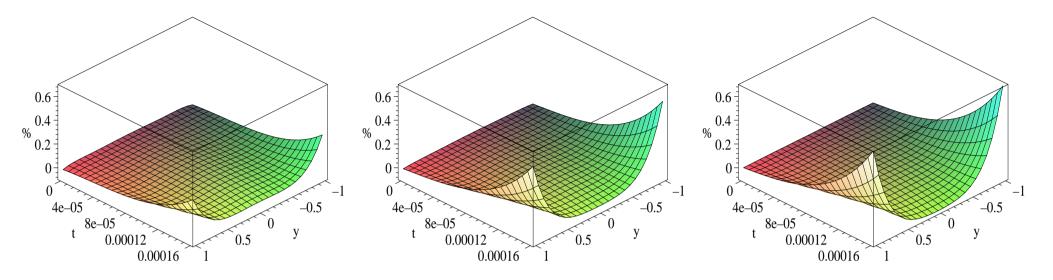
Displacement u_x ($\varepsilon = 0$)



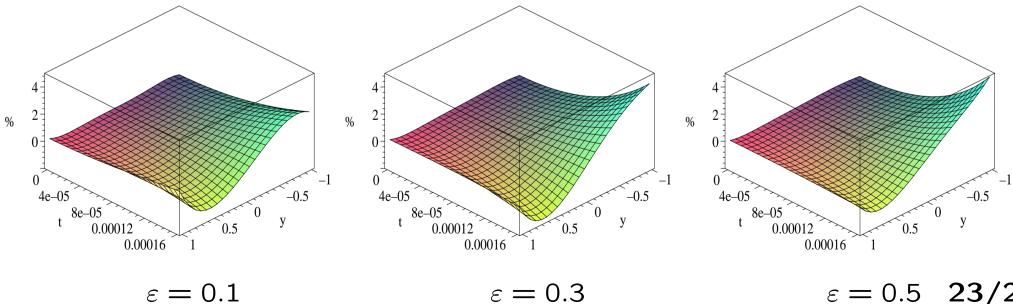


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The relative error for $u_x(1, y, t)$ (inverse problem)



The disturbed internal responses - normal distribution N(0,0.04)



 $\varepsilon = 0.5$ 23/25

The mean relative error of approximation of boundary condition $u_x(1, y, t)$ is defined as:

$$\mathsf{E} = \sqrt{\frac{\int_{-1}^{1} \int_{0}^{\Delta t} [\hat{u}_{x}(1;y;t) - u_{x}(1;y;t)]^{2} \mathrm{d}t \mathrm{d}y}{\int_{-1}^{1} \int_{0}^{\Delta t} [u_{x}(1;y;t)]^{2} \mathrm{d}t \mathrm{d}y}} \cdot 100\%.$$

In approximation \hat{u}_x we take all wave polynomials up to order 7. The table shows the error E[%] in dependence on the distance ε :

ε	Smooth data	Noisy data
0	0.022	0.896
0.1	0.044	1.073
0.3	0.108	1.503
0.5	0.133	1.394

For noisy data the error is bigger but still stays very low even for big distance ε .

Conclusions

- A new, relatively simple method of solving the direct and inverse problems for wave equations was proposed.
- Thanks to this method we obtain an analytical solution which satisfies given equation and depends continuously on all variables in the whole domain.
- The error of approximation is small.
- Considered example confirms the theoretical result that the wave polynomials method is convergent – more polynomials in approximate solution leads to better results.
- The method is flexible according to initial and boundary conditions (discrete, missing).
- A new and very effective approach towards smoothing by using Trefftz functions is proposed here.

Thank you for your attention!