

Methods for inverse problems:

VI. Using higher derivatives: Halley's method

Barbara Kaltenbacher, University of Klagenfurt, Austria

nonlinear inverse problem

$$F(x) = y$$

nonlinear inverse problem

$$F(x) = y$$

such as parameter identification in PDE

$$A(x, u) = 0 \quad y = Cu$$

nonlinear inverse problem

$$F(x) = y$$

such as parameter identification in PDE

$$A(x, u) = 0 \quad y = Cu$$

e.g., identification of space dependent diffusion a in elliptic PDE

$$\begin{cases} -\nabla(a\nabla u) = f & \text{in } \Omega \\ u = g & \text{on } \partial\Omega \end{cases} \quad y = u \text{ or } y = u|_{\omega} \text{ or } y = u|_{\partial\Omega}$$

nonlinear inverse problem

$$F(x) = y$$

such as parameter identification in PDE

$$A(x, u) = 0 \quad y = Cu$$

e.g., identification of space dependent diffusion a in elliptic PDE

$$\begin{cases} -\nabla(a\nabla u) = f & \text{in } \Omega \\ u = g & \text{on } \partial\Omega \end{cases} \quad y = u \text{ or } y = u|_{\omega} \text{ or } y = u|_{\partial\Omega}$$

e.g., identification of space dependent potential c in elliptic PDE

$$\begin{cases} -\Delta u + cu = f & \text{in } \Omega \\ u = g & \text{on } \partial\Omega \end{cases} \quad y = u \text{ or } y = u|_{\omega} \text{ or } y = u|_{\partial\Omega}$$

Newton's method relies on 1st order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k)$$

Newton's method relies on 1st order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k)$$

$$x_{k+1} = x_k + F'(x_k)^{-1}(y - F(x_k))$$

Newton's method relies on 1st order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k)$$

$$x_{k+1} = x_k + F'(x_k)^{-1}(y - F(x_k))$$

F smooth \rightsquigarrow 2nd order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k) + \frac{1}{2}F''(x_k)(x - x_k)^2$$

Newton's method relies on 1st order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k)$$

$$x_{k+1} = x_k + F'(x_k)^{-1}(y - F(x_k))$$

F smooth \rightsquigarrow 2nd order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k) + \frac{1}{2}F''(x_k)(x - x_k)^2$$

$$x_{k+1} = ? \quad \text{quadratic operator equation}$$

Newton's method relies on 1st order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k)$$

$$x_{k+1} = x_k + F'(x_k)^{-1}(y - F(x_k))$$

F smooth \rightsquigarrow 2nd order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k) + \frac{1}{2}F''(x_k)(x - x_k)^2$$

$$x_{k+1} = ? \quad \text{quadratic operator equation}$$

alternative: intermediate step x_{k+} and Taylor expansion of F' :

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \quad \text{Newton step}$$

$$F'(x_{k+}) \approx F'(x_k) + F''(x_k)(x_{k+} - x_k) =: S_k$$

$$x_{k+1} = x_k + S_k^{-1}(y - F(x_k)) \quad \text{enhanced Newton step}$$

Newton's method relies on 1st order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k)$$

$$x_{k+1} = x_k + F'(x_k)^{-1}(y - F(x_k))$$

F smooth \rightsquigarrow 2nd order Taylor expansion:

$$y = F(x) \approx F(x_k) + F'(x_k)(x - x_k) + \frac{1}{2}F''(x_k)(x - x_k)^2$$

$$x_{k+1} = ? \quad \text{quadratic operator equation}$$

alternative: intermediate step x_{k+} and Taylor expansion of F' :

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \quad \text{Newton step}$$

$$F'(x_{k+}) \approx F'(x_k) + F''(x_k)(x_{k+} - x_k) =: S_k$$

$$x_{k+1} = x_k + S_k^{-1}(y - F(x_k)) \quad \text{enhanced Newton step}$$

uses F'' , converges cubically in well-posed case \rightsquigarrow 2nd order method

Halley's method, method of tangent hyperbolas

[Brown'77], [Döring'70], [Ren&Argyros'12] well-posed

[Hettlich&Rundell'00] ill-posed problems

Why not just do a second Newton step?

two Newton steps:

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \text{ 1st Newton step}$$

$$x_{k+1} = x_{k+} + F'(x_{k+})^{-1}(y - F(x_{k+})) \text{ 2nd Newton step}$$

intermediate step x_{k+} and Taylor expansion of F' :

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \text{ 1st Newton step}$$

$$F'(x_{k+}) \approx F'(x_k) + F''(x_k)(x_{k+} - x_k, \cdot) =: S_k$$

$$x_{k+1} = x_k + S_k^{-1}(y - F(x_k)) \text{ enhanced Newton step}$$

two Newton steps:

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \text{ 1st Newton step}$$

$$x_{k+1} = x_{k+} + F'(x_{k+})^{-1}(y - F(x_{k+})) \text{ 2nd Newton step}$$

intermediate step x_{k+} and Taylor expansion of F' :

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \text{ 1st Newton step}$$

$$F'(x_{k+}) \approx F'(x_k) + F''(x_k)(x_{k+} - x_k, \cdot) =: S_k$$

$$x_{k+1} = x_k + S_k^{-1}(y - F(x_k)) \text{ enhanced Newton step}$$

two Newton steps:

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \text{ 1st Newton step}$$

$$x_{k+1} = x_{k+} + F'(x_{k+})^{-1}(y - F(x_{k+})) \text{ 2nd Newton step}$$

intermediate step x_{k+} and Taylor expansion of F' :

$$x_{k+} = x_k + F'(x_k)^{-1}(y - F(x_k)) \text{ 1st Newton step}$$

$$F'(x_{k+}) \approx F'(x_k) + F''(x_k)(x_{k+} - x_k, \cdot) =: S_k$$

$$x_{k+1} = x_k + S_k^{-1}(y - F(x_k)) \text{ enhanced Newton step}$$

After 1st Newton step $F''(x_k)(x_{k+} - x_k, \cdot)$ is cheaper to evaluate than $F'(x_{k+})$, $F(x_{k+})$!

e.g., parameter identification in PDEs:

$$F(x_k) = CS(x_k) \text{ with } S(x_k) = u \text{ satisfying } A(x_k, u) = 0$$

derivatives:

$$F'(x_k)h = C\tilde{u}(h) \quad \text{with } \tilde{u}(h) \text{ satisfying } A_u(x_k, u)\tilde{u} = f_1(h)$$

$$F''(x_k)(x_{k+} - x_k, h) = C\tilde{\tilde{u}}(h) \quad \text{with } \tilde{\tilde{u}}(h) \text{ satisfying } A_u(x_k, u)\tilde{\tilde{u}} = f_2(h)$$

with

$$f_1(h) = -A_x(x_k, u)h$$

$$f_2(h) = -A_{xx}(x_k, u)(x_{k+} - x_k, h) - A_{xu}(x, u)(h, \tilde{u}(x_{k+} - x_k)) \\ - A_{ux}(x, u)(\tilde{u}(x_{k+} - x_k), h) - A_{uu}(x, u)(\tilde{u}(x_{k+} - x_k), \tilde{u})$$

Same PDE with same parameter for $F'(x_k)$ and $F''(x_k)(x_{k+} - x_k)$
with different right hand sides

e.g., identification of diffusion in elliptic PDE

$$\begin{aligned} -\nabla(a\nabla u) &= f && \text{in } \Omega \\ u &= g && \text{on } \partial\Omega \end{aligned}$$

from measurements $y = Cu$ of u .

$$F'(a)h = CS'(a)h, \quad F''(a)(h, \ell) = CS''(a)(h, \ell)$$

with $\tilde{u} = S'(a)h$, $\tilde{\tilde{u}} = S''(a)(h, \ell)$ defined by

$$\begin{aligned} -\nabla(a\nabla\tilde{u}) &= \nabla(h\nabla S(a)) && \text{in } \Omega \\ \tilde{u} &= 0 && \text{on } \partial\Omega \end{aligned}$$

$$\begin{aligned} -\nabla(a\nabla\tilde{\tilde{u}}) &= \nabla(h\nabla S'(a)\ell) + \nabla(\ell\nabla S'(a)h) && \text{in } \Omega \\ \tilde{\tilde{u}} &= 0 && \text{on } \partial\Omega \end{aligned}$$

Systems for $F(a)$, $F'(a)$, $F''(a)$ contain the same stiffness matrix.
Evaluation of $F'(a_+)$ would require new stiffness matrix.

e.g., identification of potential in elliptic PDE

$$\begin{aligned} -\Delta u + cu &= f & \text{in } \Omega \\ u &= g & \text{on } \partial\Omega \end{aligned}$$

from measurements $y = Cu$ of u .

$$\begin{aligned} F'(c)h &= CS'(c)h, \quad F''(c)(h, \ell) = CS''(c)(h, \ell) \\ \text{with } \tilde{u} &= S'(c)h, \quad \tilde{\tilde{u}} = S''(c)(h, \ell) \end{aligned}$$

$$\begin{aligned} -\Delta \tilde{u} + c\tilde{u} &= hS(c) & \text{in } \Omega \\ \tilde{u} &= 0 & \text{on } \partial\Omega \end{aligned}$$

$$\begin{aligned} -\Delta \tilde{\tilde{u}} + c\tilde{\tilde{u}} &= hS'(c)\ell + \ell S'(c)h & \text{in } \Omega \\ \tilde{\tilde{u}} &= 0 & \text{on } \partial\Omega \end{aligned}$$

Systems for $F(c)$, $F'(c)$, $F''(c)$ contain the same stiffness matrix.
Evaluation of $F'(c_+)$ would require new stiffness matrix.

Halley's method for ill-posed problems

Halley's method for ill-posed problems in Hilbert spaces

$$T_k = F'(x_k^\delta); \quad r_k = F(x_k^\delta) - y^\delta$$

$$x_{k+}^\delta = x_k^\delta - (T_k^* T_k + \beta_k I)^{-1} \{T_k^* r_k + \beta_k (x_k^\delta - x_0)\}$$

$$S_k = T_k + \frac{1}{2} F''(x_k^\delta)(x_{k+}^\delta - x_k^\delta, \cdot)$$

$$x_{k+1}^\delta = x_k - (S_k^* S_k + \alpha_k I)^{-1} \{S_k^* r_k + \alpha_k (x_k^\delta - x_0)\}$$

with a priori fixed sequences of regularization parameters $(\alpha_k)_{k \in \mathbb{N}}$, $(\beta_k)_{k \in \mathbb{N}}$ satisfying

$$\alpha_k \searrow 0, \quad \beta_k \searrow 0, \quad 1 \leq \frac{\alpha_k}{\alpha_{k+1}} \leq q, \quad 1 \leq \frac{\beta_k}{\beta_{k+1}} \leq q,$$

and a priori stopping rule $k_* = k_*(\delta)$ depending on noise level

$$\delta \geq \|y^\delta - y\|.$$

[Hettlich&Rundell'00]: Levenberg-Marquardt type
a posteriori regularization parameter choice,
Hilbert space setting, convergence without rates

Halley's method for ill-posed problems in Banach spaces

$$T_k = F'(x_k^\delta); \quad r_k = F(x_k^\delta) - y^\delta$$

$$x_{k+}^\delta \in \operatorname{argmin}_x \frac{1}{r} \|T_k(x - x_k^\delta) + r_k\|^r + \frac{\beta_k}{p} \|x - x_0\|^p$$

$$S_k = T_k + \frac{1}{2} F''(x_k^\delta)(x_{k+}^\delta - x_k^\delta, \cdot)$$

$$x_{k+1}^\delta \in \operatorname{argmin}_x \frac{1}{r} \|S_k(x - x_k^\delta) + r_k\|^r + \frac{\alpha_k}{p} \|x - x_k^\delta\|^p$$

with $p, r \in [1, \infty)$

a priori fixed sequences of regularization parameters $(\alpha_k)_{k \in \mathbb{N}}$,
 $(\beta_k)_{k \in \mathbb{N}}$ satisfying

$$\alpha_k \searrow 0, \quad \beta_k \searrow 0, \quad 1 \leq \frac{\alpha_k}{\alpha_{k+1}} \leq q, \quad 1 \leq \frac{\beta_k}{\beta_{k+1}} \leq q, \quad (1)$$

and a priori stopping rule $k_* = k_*(\delta)$ depending on noise level

$$\delta \geq \|y^\delta - y\|.$$

convergence results

Theorem

Assume that a source condition

$$x^\dagger - x_0 = (T^* T)^\mu v$$

with $\mu \in [\frac{1}{2}, 1]$, and $\|v\|$ sufficiently small holds.

Let F'' be bounded and Lipschitz continuous in a neighborhood of x^\dagger and let x_0 be sufficiently close to x^\dagger .

Assume that $\beta_k = \alpha_k$ is chosen so that (1) holds and let k_* be chosen as the first index such that

$$\alpha_{k_*}^{\mu + \frac{1}{2}} \leq \tau \delta$$

with τ sufficiently large.

Then

$$\|x_{k_*}^\delta - x^\dagger\| = O(\delta^{\frac{2\mu}{2\mu+1}}) \text{ as } \delta \rightarrow 0.$$

If $\delta = 0$

$$\|x_k^\delta - x^\dagger\| = O(\alpha_k^\mu) \text{ as } k \rightarrow \infty.$$

Convergence with weak or no regularity of $x^\dagger - x_0$

source condition

$$x^\dagger - x_0 = f(T^*T)v \quad (2)$$

with $f : (0, \infty) \rightarrow (0, \infty)$ continuous and strictly increasing with

$$f(\lambda) \rightarrow 0, \quad \frac{\lambda}{f(\lambda)} \leq C \text{ as } \lambda \rightarrow 0, \quad \mu_f := \sup_{\alpha \in (0, \alpha_0]} \frac{f'(\alpha)\alpha}{f(\alpha)} < \infty \quad (3)$$

e.g., $f(\lambda) = \log(1/\lambda)^p$, $f(\lambda) = \lambda^\mu$, $\mu \leq 1$

or no source condition

$$x^\dagger - x_0 \in \mathcal{N}(T)^\perp \quad (4)$$

a priori stopping rule: k_* is chosen as the first index such that

$$\sqrt{\alpha_{k_*}} f(\alpha_{k_*}) \leq \tau \delta$$

with τ sufficiently large and – if we only have (know) (4) – just

$$k_* \rightarrow \infty \quad \frac{\delta}{\sqrt{\alpha_{k_*}}} \rightarrow 0 \text{ as } \delta \rightarrow 0 \quad (5)$$

Structural condition on F : range invariance condition

$F'(\tilde{x}) = F'(x)R(x, \tilde{x})$ with

$R(x, \tilde{x}) \in \mathcal{L}(X, X)$, $\|R(\tilde{x}, x) - I\| \leq M_1 \|\tilde{x} - x\| \quad \forall x, \tilde{x} \in \mathcal{B}_\rho(x^\dagger)$

$F''(x)(h, l) = F'(x)R_2^x(h, l)$ with

$R_2^x(h, \cdot)$, $R_2^x(\cdot, l) \in \mathcal{L}(X, X)$, $\|R_2^x\| \leq M_2 \quad \forall x \in \mathcal{B}_\rho(x^\dagger)$, $h, l \in X$
(6)

e.g., diffusion identification from (complete, partial, boundary)
measurements of u in 1-d

e.g., potential identification from (complete, partial, boundary)
measurements of u in 3-d

Theorem

Let F satisfy the range invariance condition (6) and let x_0 be sufficiently close to x^\dagger and satisfy (4). Assume that $\beta_k = \alpha_k$, is chosen so that (1) holds and let k_* be chosen according to (5). Then the iterates $x_{k_*}^\delta$ converge to x^\dagger as $\delta \rightarrow 0$. If a source condition (2) with (3) is satisfied, then the rate

$$\|x_{k_*}^\delta - x^\dagger\| = O(f(\Theta^{-1}(\delta))) = O\left(\frac{\delta}{\sqrt{\Theta^{-1}(\delta)}}\right) \text{ as } \delta \rightarrow 0.$$

is obtained, where $\Theta(\lambda) := f(\lambda)\sqrt{\lambda}$. If $\delta = 0$ we have convergence

$$\|x_k^\delta - x^\dagger\| = O(f(\alpha_k)) \text{ as } k \rightarrow \infty.$$

numerical results

Test problem: Potential identification

Identify c in

$$\begin{aligned} -\Delta u + \Phi(c)u &= f && \text{in } \Omega \\ u &= g && \text{on } \partial\Omega \end{aligned}$$

from measurements $y = Cu$ of u ,

where $\Phi(\lambda) = \frac{1}{2}\lambda^2\mathbb{I}_{[-\bar{c},\bar{c}]} + \frac{1}{2}\bar{c}(2|\lambda| - \bar{c})\mathbb{I}_{\mathbb{R}\setminus[-\bar{c},\bar{c}]}$, so that

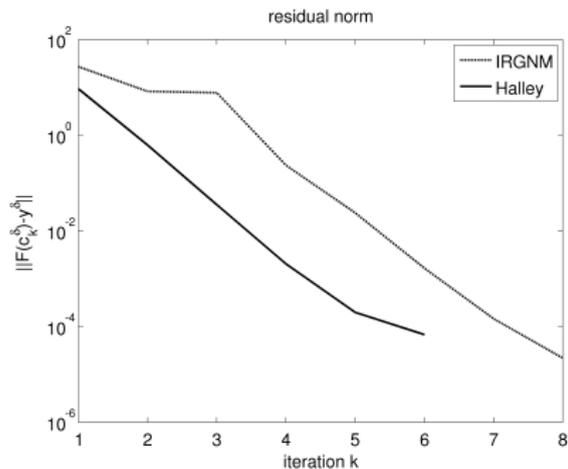
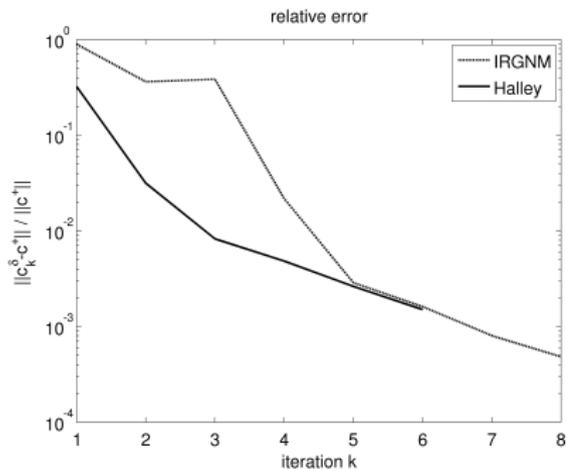
$$c \in L^2(\Omega) \quad \Rightarrow \quad \Phi(c) \geq 0 \text{ and } \Phi(c) \in L^2(\Omega)$$

$$\Omega = (0, 1)^2$$

$$c(x_1, x_2) = 1 + \frac{1}{2}\xi(1 - \cos(4\pi x_1))(1 - \cos(4\pi x_2))\mathbb{I}_{(0, \frac{1}{2})^2}$$

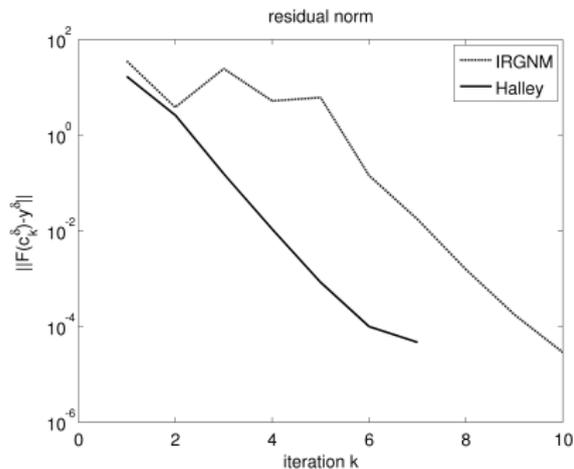
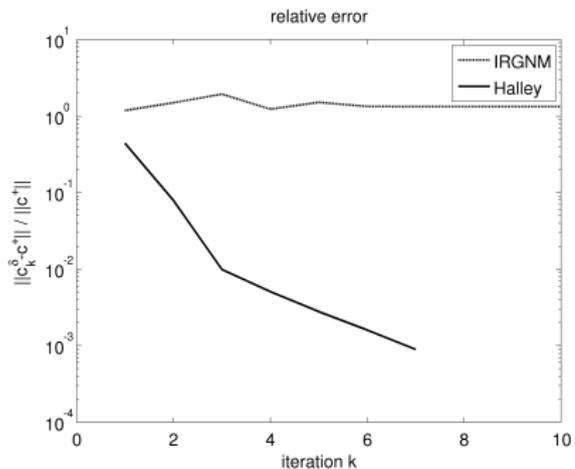
with $\xi \in \{5, 7, 10\}$,
starting value $c_0 \equiv 1$.

Comparison of IRGNM (dashed) and Halley (solid)



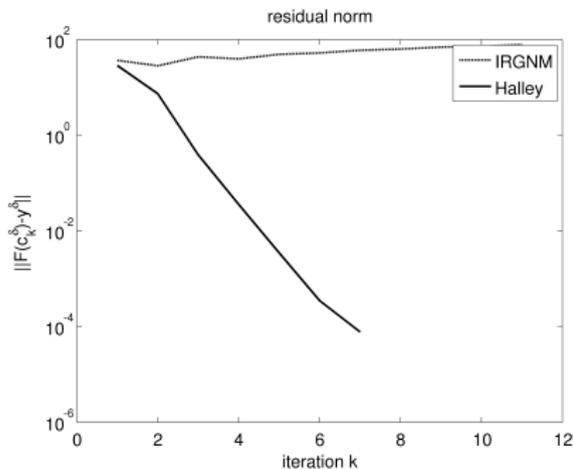
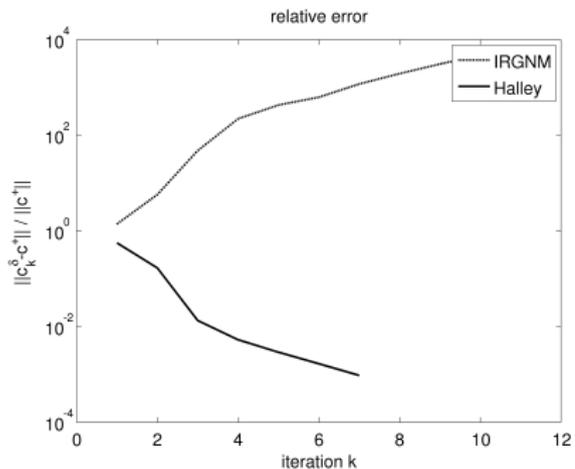
relative error (left) and residual (right) for $\xi = 5$

Comparison of IRGNM (dashed) and Halley (solid)



relative error (left) and residual (right) for $\xi = 7$

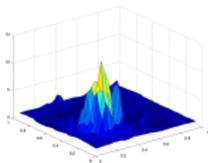
Comparison of IRGNM (dashed) and Halley (solid)



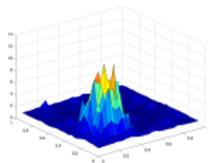
relative error (left) and residual (right) for $\xi = 10$

Reconstructions (top) from data (bottom) with Gaussian noise of decreasing level

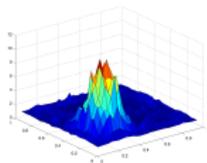
$\delta = 1\%$



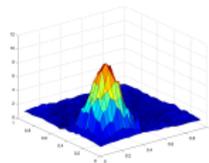
$\delta = 0.5\%$



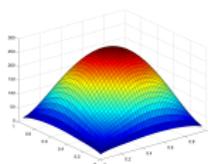
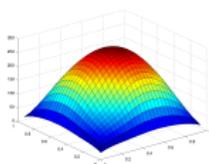
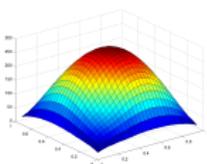
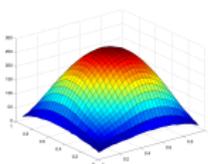
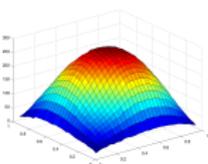
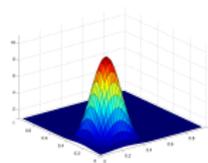
$\delta = 0.25\%$



$\delta = 0.125\%$

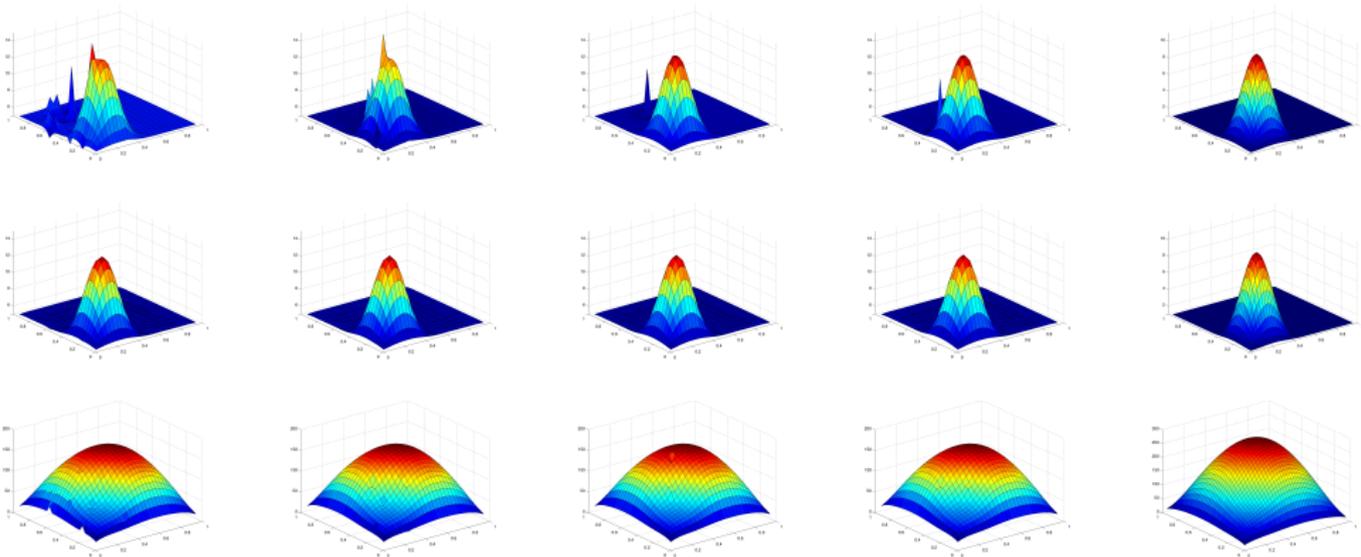


$\delta = 0\%$



$\xi = 10$

Reconstructions with $Y = L^2$ (top) and $Y = L^{1.1}$ (middle)
from data (bottom) with impulsive noise of decreasing
amount from left to right



Conclusions and Outlook

- higher order methods seem to pay off in parameter identification for PDEs
 - existing analysis:
 - convergence (rates) in high and low regularity (source condition) case
 - convergence rate under benchmark source condition in Banach spaces
- several open questions in analysis (rates with a posteriori regularization parameter choice, general rates in Banach spaces, . . .)
- n stage versions of Halley's method

$$T_k^0 = 0 \quad r_k = F(x_k^\delta) - y^\delta$$

for $j = 1, \dots, n$ do

$$T_k^j = T_k^{j-1} + \sum_{m=1}^j \frac{1}{m!} F^{(m)}(x_k^\delta) ((x_{k+\frac{m-1}{n}}^\delta - x_k^\delta)^{m-1}, \cdot)$$

$$x_{k+\frac{j}{n}}^\delta = x_k^\delta - (T_k^{j*} T_k^j + \alpha_k^j I)^{-1} \left\{ T_k^{j*} r_k + \alpha_k^j (x_k^\delta - x_0) \right\}$$