

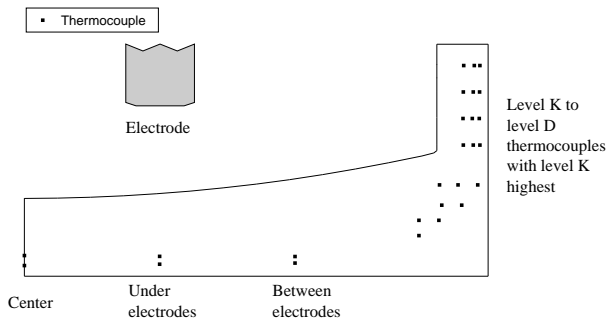
Solving Ill-Posed Cauchy Problems using Rational Krylov Methods

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GAMM 2008

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Motivating example: Ilmenite iron melting furnace



The furnace material properties are temperature dependent.

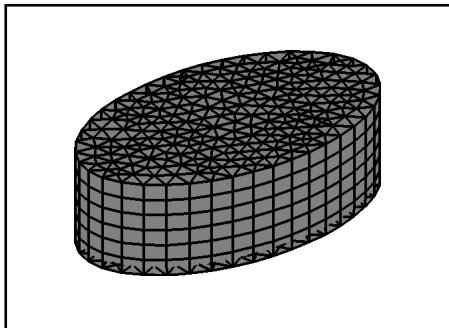
Problem: **find the inner shape of the furnace.**

Nonlinear, and (rather) complex geometry

PhD thesis: I M Skaar, Monitoring the Lining of a Melting Furnace, NTNU, Trondheim, 2001

Inverse Heat Conduction Problem

Steady state heat conduction problem:



The upper boundary is unavailable for measurements

Ill-Posed Cauchy Problem

Ω : connected in \mathbb{R}^2 with smooth boundary $\partial\Omega$

L : linear, self-adjoint, positive definite elliptic in Ω .

$$u_{zz} - Lu = 0, \quad (x, y) \in \Omega, \quad z \in [0, z_1],$$

$$u(x, y, z) = 0, \quad (x, y) \in \partial\Omega, \quad z \in [0, z_1],$$

$$u(x, y, 0) = g(x, y), \quad (x, y) \in \Omega,$$

$$u_z(x, y, 0) = 0, \quad (x, y) \in \Omega.$$

Sought: $f(x, y) = u(x, y, z_1)$, $(x, y) \in \Omega$.

Formal solution

$$u(x, y, z) = \cosh(z\sqrt{L})g$$

BUT: L is a pos. def. unbounded operator \Rightarrow ILL-POSED!

Standard Iterative Procedure

Guess $f^{(1)}$

1 for $k = 1, 2, \dots$ until convergence

1 Solve

$$\begin{aligned}u_{zz} - Lu &= 0, & (x, y) \in \Omega, \quad z \in [0, z_1], \\u(x, y, z) &= 0, & (x, y) \in \partial\Omega, \quad z \in [0, z_1], \\u(x, y, z_1) &= f^{(k)}, & (x, y) \in \Omega, \\u_z(x, y, 0) &= 0, & (x, y) \in \Omega.\end{aligned}$$

giving $u^{(k)}$

2 Evaluate $\|g(\cdot, \cdot) - u^{(k)}(\cdot, \cdot, 0)\|$ and adjust $f^{(k)} \rightarrow f^{(k+1)}$

In every iteration: **Solve a 3D well-posed problem**

Often slow convergence.

Other possible methods?

- Tikhonov regularization?

Impossible, because we do not know the integral operator for equations with variable coefficients and/or complicated geometry.

- Replace unbounded L by a bounded approximation?

Possible in connection with finite difference approximation, but more difficult with finite elements.

BUT: Krylov method!

Regularization

Formal solution

$$u(x, y, z) = \cosh(z\sqrt{L})g$$

BUT: L is a pos. def. unbounded operator \Rightarrow **ILL-POSED!**

High frequency perturbations in g are blown up

Regularization

Replace unbounded operator by a bounded one!

Cut Off High Frequencies

Eigenvalues of L : $\lambda_j^2, j = 1, 2, \dots$ and $\lambda_j \rightarrow +\infty$ as $j \rightarrow \infty$

General approach: Compute the k eigenvalues of smallest modulus:

$$LX_k = X_k D_k,$$

where X_k holds orthonormal eigenvectors

Approximate by projection

$$\cosh(z\sqrt{L})g \approx \cosh(z\sqrt{L})X_k X_k^\top g = X_k \cosh(z\sqrt{D_k})X_k^\top g$$

Eigenvalues of L

L is *large and sparse* (of the order $10^4 - 10^5$, say)

Compute the **smallest** eigenvalues



Operate with L^{-1}



Solve many standard 2D elliptic problems

$$-Lw = v$$

$L_2(\Omega)$ setting, u is an “exact solution”

v is an approximate solution with perturbed data g_m

Theorem

Assume that $\|u(\cdot, \cdot, 1)\| \leq M$ and that the data perturbation satisfies $\|g - g_m\| \leq \epsilon$. Then if v is computed by projection using the eigenvalues satisfying $\lambda_j \leq \lambda_c$, where

$$\lambda_c = (1/z_1) \log(M/\epsilon)$$

then

$$\|u(\cdot, \cdot, z) - v(\cdot, \cdot, z)\| \leq 3\epsilon^{1-z/z_1} M^{z/z_1}, \quad 0 \leq z \leq z_1.$$

Optimal error bound

- Is it necessary to compute the eigenvalues and eigenvectors accurately?
- Do we need all the information that we get in the eigenvalues?
- Can we take advantage of the fact that we want to compute an approximation of

$$\cosh(z\sqrt{L})g$$

for **this particular vector**?

- Is it necessary to compute the eigenvalues and eigenvectors accurately? **NO!**
- Do we need all the information that we get in the eigenvalues? **NO!**
- Can we take advantage of the fact that we want to compute an approximation of

$$\cosh(z\sqrt{L})g$$

for **this particular vector?** **YES!**

Lanczos tridiagonalization

Choose q_1 and iterate

$$L^{-1}q_k = q_{k-1}\beta_{k-1} + q_k\alpha_k + q_{k+1}\beta_k, \quad k = 1, 2, \dots,$$

with $\alpha_k = q_k^\top L^{-1}q_k$ and $\beta_k = q_{k+1}^\top L^{-1}q_k$;

One matrix-vector multiply $L^{-1}q_k$ per step

One standard 2D elliptic solve (black box) per step

$$-Lw = q_k$$

- Initial convergence influenced by starting vector v_1 . Choose $q_1 = 1/\beta g_m$, $\beta = \|g_m\|$
- Faster for largest eigenvalues of $L^{-1} \Rightarrow$ Fast convergence for **some of the smallest eigenvalues** of L
- Optional: To get faster convergence for eigenvalues in $[0, \lambda_c]$ operate with $(L - \tau I)^{-1}$, where $\tau = \lambda_c/2$

$$L^{-1} Q_k = Q_k T_k + \beta_{k+1} q_{k+1} e_{k+1}^\top \approx Q_k T_k$$

Approximation

$$\cosh(z\sqrt{L})g \approx Q_k \cosh(zT_k^{-1/2})Q_k^\top g$$

Problem: We cannot prevent T_k from approximating large eigenvalues!

Solution: Regularize T_k : cut off large eigenvalues¹

¹**Krylov+regularization:** O'Leary & Simmons (1981), Björck, Grimme & Van Dooren (1994)

Projected and Truncated Approximation

Let

$$((\theta_j^{(k)})^2, y_j^{(k)}), j = 1, \dots, k$$

be the eigenpairs of T_k^{-1}

Define $F(z, \lambda) = \cosh(z\lambda^{1/2})$ and $S_k = T_k^{-1}$

Truncated approximation:

$$\begin{aligned} v_k(z) &= Q_k F(z, S_k^c) Q_k^\top g_m \\ &:= Q_k \sum_{\theta_j^{(k)} \leq \lambda_c} y_j^{(k)} \cosh(z\theta_j^{(k)}) (y_j^{(k)})^\top e_1 \|g_m\|. \end{aligned}$$

Recall $F(z, \lambda) = \cosh(z\lambda^{1/2})$ and $S_k = T_k^{-1}$

Theorem

Let u be the “exact solution” and

$$v_k(z) = Q_k F(z, S_k^c) Q_k^T g_m$$

Under the same hypotheses as earlier,

$$\begin{aligned} \|u(z) - v_k(z)\| &\leq 3\epsilon^{1-z/z_1} M^{z/z_1} \\ &+ 2\|[F(z, L^c) - Q_k F(z, S_k^c) Q_k^T]g\|. \end{aligned}$$

- Starting vector $q_1 = 1/\beta g_m$
- **for** $k = 2, 3, \dots$ **until** “stable”
 - $[Q_k, T_k] = \text{krylovstep}(L^{-1}, Q_{k-1}, T_{k-1})$
 - Compute $v_k(z) = Q_k F(z, S_k^c) Q_k^\top g_m$
- **end**
- Check residual $\|Kv_k - g_m\| < \epsilon$

Kv_k is the solution of the 3D problem with $u = v_k$ at the upper boundary and $u_z = 0$ at the lower. **Expensive!**

Residual: $\|Kv_k - g_m\| < \epsilon$

Solve 3D problem (denote solution u_k)

$$\begin{aligned}u_{zz} - Lu &= 0, & (x, y) \in \Omega, \quad z \in [0, z_1], \\u(x, y, z) &= 0, & (x, y) \in \partial\Omega, \quad z \in [0, z_1], \\u(x, y, 1) &= v_k(x, y), & (x, y) \in \Omega, \\u_z(x, y, 0) &= 0, & (x, y) \in \Omega.\end{aligned}$$

Well-posed but **expensive!**

$$Kv_k = u_k(x, y, 0)$$

We only want to compute this when we are sure that $\|Kv_k - g_m\| < \epsilon$

- Starting vector $q_1 = 1/\beta g_m$
- **for** $k = 1, 2, \dots$ **until** “stable”
 - $[Q_k, T_k] = \text{krylovstep}(L^{-1}, Q_{k-1}, T_{k-1})$
 - Compute $v_k(z) = Q_k F(z, S_k^c) Q_k^T g_m$
- **end**
- Check residual $\|Kv_k - g_m\| < \epsilon$

How can we quantify “stable”?

Error Estimate for Krylov Procedure

Recall $F(z, \lambda) = \cosh(z\lambda^{1/2})$ and $S_k = T_k^{-1}$

$$\begin{aligned}\|u(z) - v_k(z)\| &\leq 3\epsilon^{1-z/z_1} M^{z/z_1} \\ &+ 2\|[F(z, L^c) - Q_k F(z, S_k^c) Q_k^T]g\|.\end{aligned}$$

Second term: Krylov approximation error for low frequency operator
applied to the exact right hand side!

Heuristic

When the approximate solution does not change much between successive Krylov steps, then this error is small.

Recall $F(z, \lambda) = \cosh(z\lambda^{1/2})$ and $S_k = T_k^{-1}$

- Starting vector $q_1 = 1/\beta g_m$
- **for** $k = 1, 2, \dots$ **maxit**
 - $[Q_k, T_k] = \text{krylovstep}(L^{-1}, Q_{k-1}, T_{k-1})$
 - Compute $w_k(z) = F(z, S_k^c) Q_k^T g_m$
 - **if** $\|w_k - w_{k-1}\| < \text{tol}$ **then**
 - **if** $\|Kv_k - g_m\| < \epsilon$ **then stop iterating**
 - **endif**
- **end**
- Compute $v_k = Q_k w_k$

Test example 1: Laplace equation

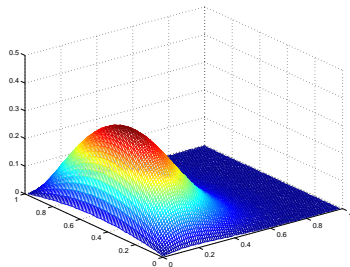
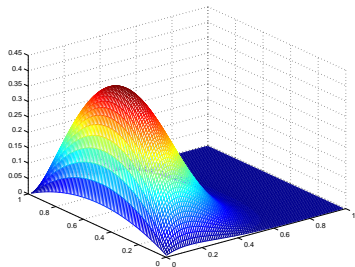
Ω : unit square

$$\begin{aligned}u_{zz} + \Delta u &= 0, & (x, y, z) \in \Omega \times [0, 0.1], \\u(x, y, z) &= 0, & (x, y, z) \in \partial\Omega \times [0, 0.1], \\u(x, y, 0) &= g(x, y), & (x, y) \in \Omega, \\u_z(x, y, 0) &= 0, & (x, y) \in \Omega.\end{aligned}$$

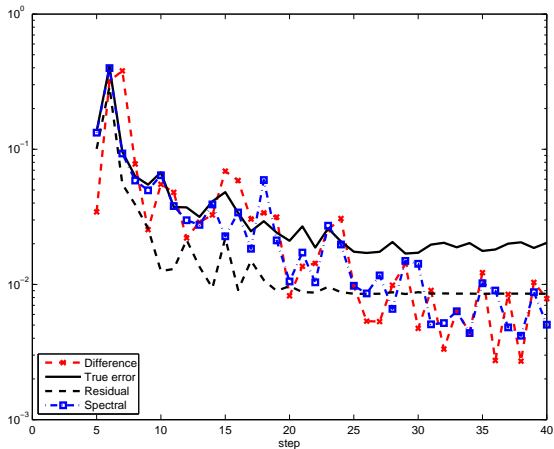
Determine the values at the upper boundary,
 $f(x, y) = u(x, y, 0.1)$, $(x, y) \in \Omega$.

Data perturbation: $\|g - g_m\|/\|g\| \approx 0.0085$
98 eigenvalues are smaller than the tolerance
`eigs` performs approximately 300 2D elliptic solves.

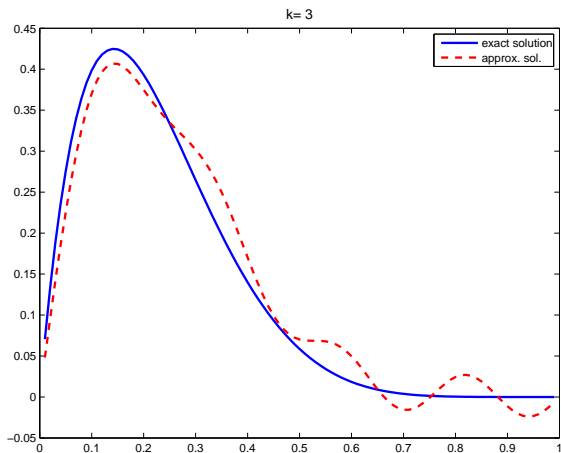
Solution and Exact Data



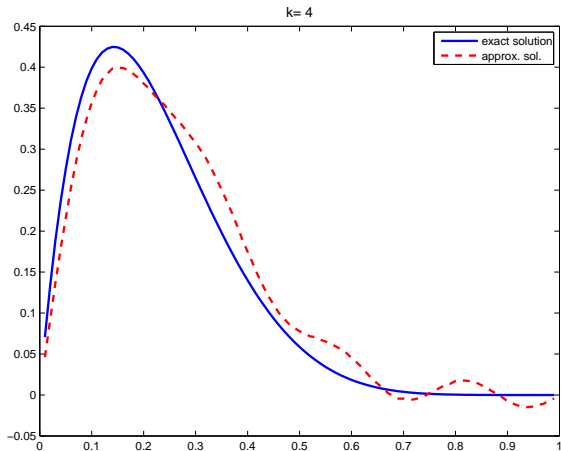
Convergence History ($0.75\lambda_c$)



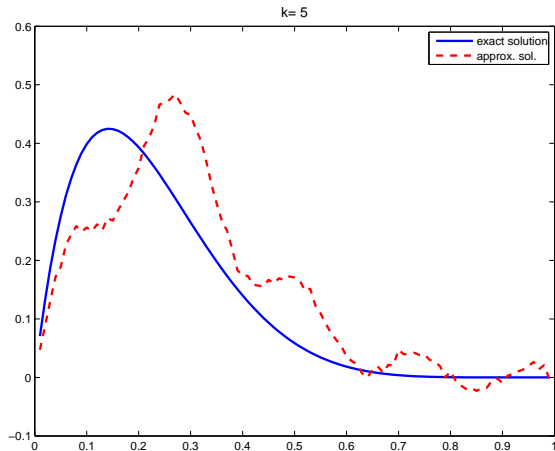
Solutions as function of iteration index



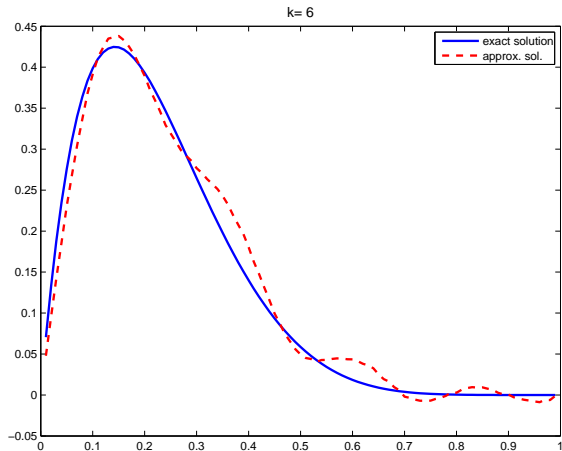
Solutions as function of iteration index



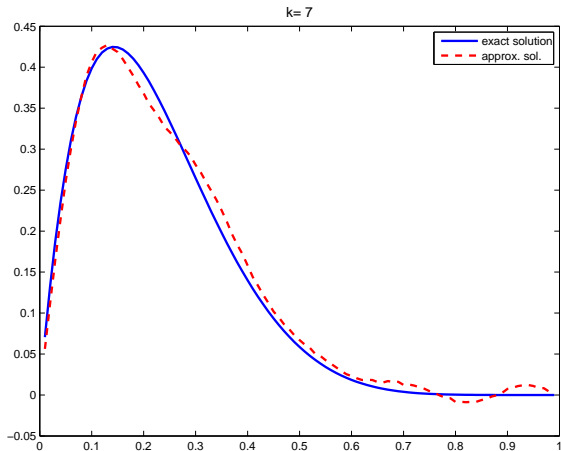
Solutions as function of iteration index



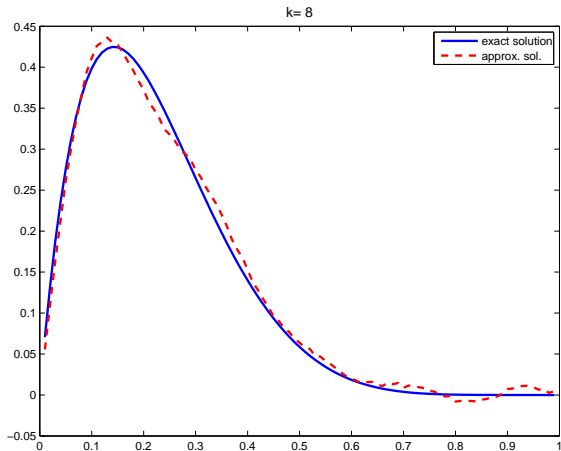
Solutions as function of iteration index



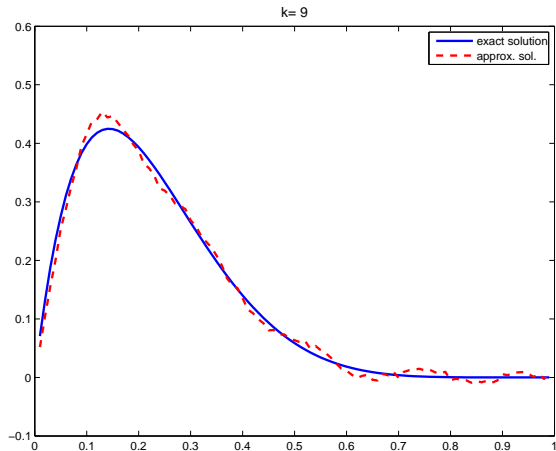
Solutions as function of iteration index



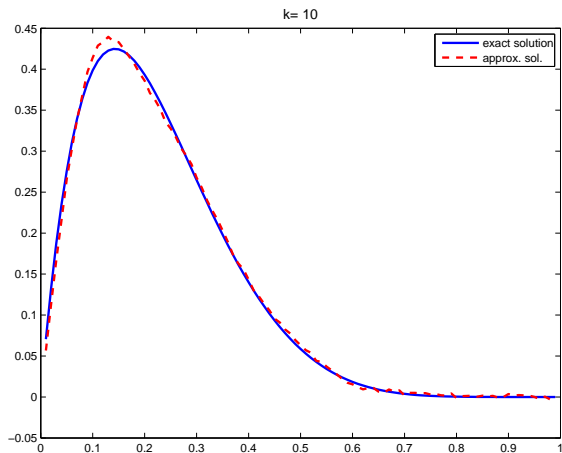
Solutions as function of iteration index



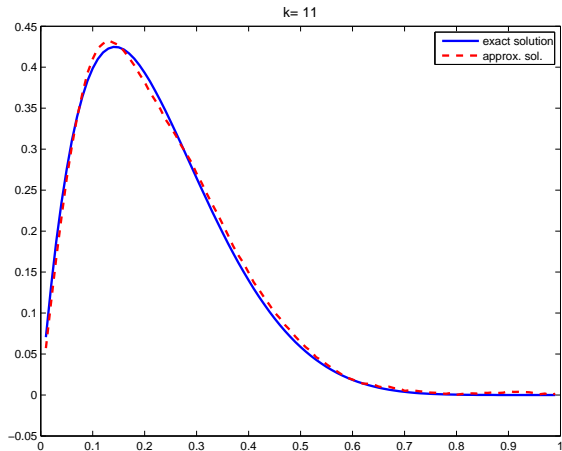
Solutions as function of iteration index



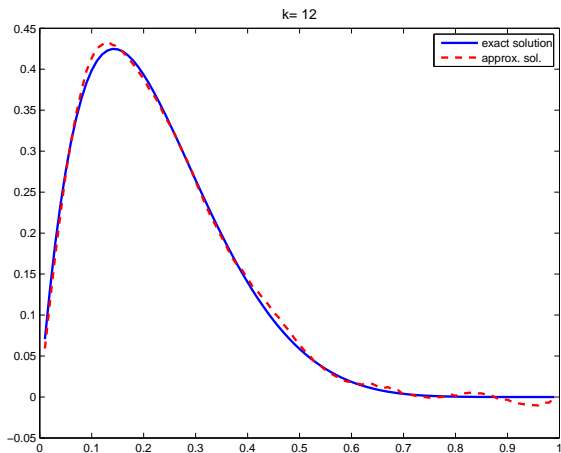
Solutions as function of iteration index



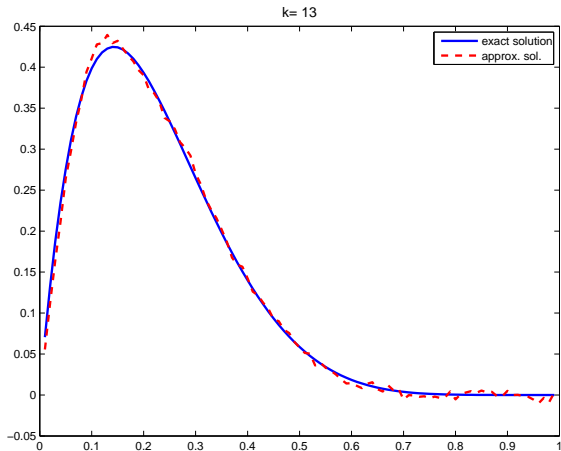
Solutions as function of iteration index



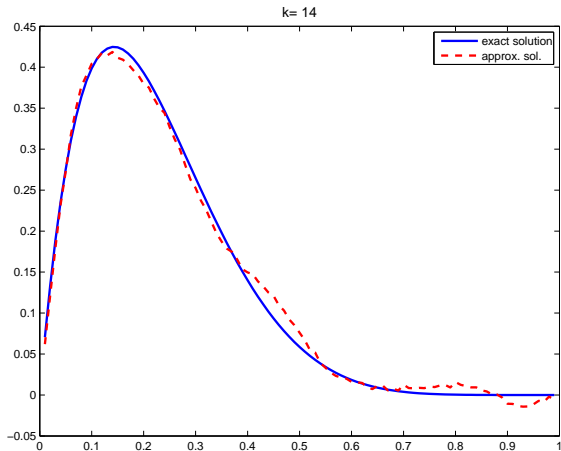
Solutions as function of iteration index



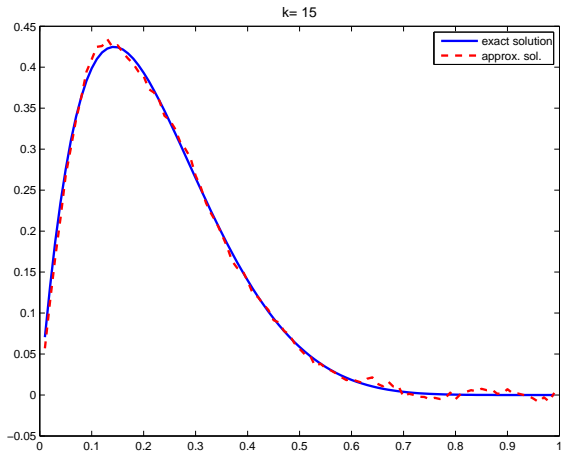
Solutions as function of iteration index



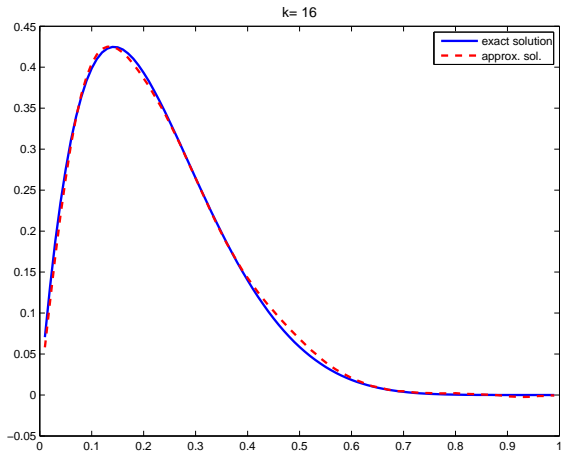
Solutions as function of iteration index



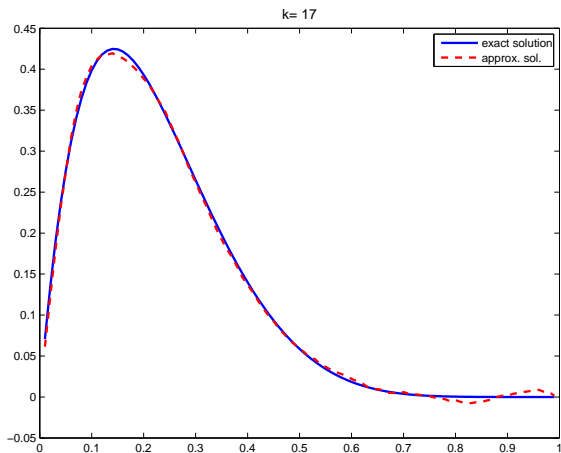
Solutions as function of iteration index



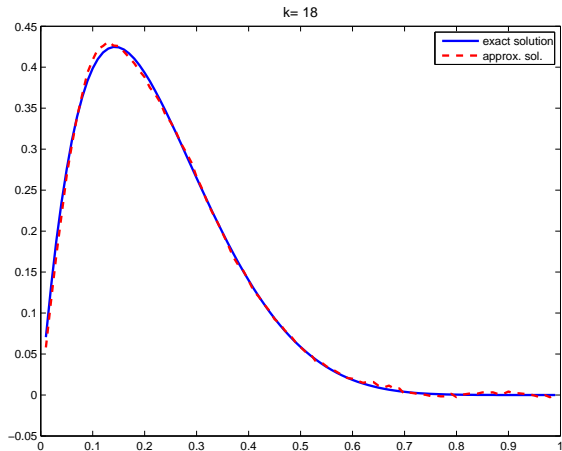
Solutions as function of iteration index



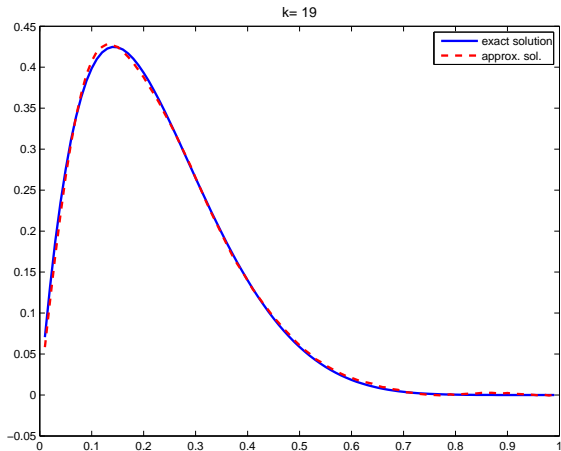
Solutions as function of iteration index



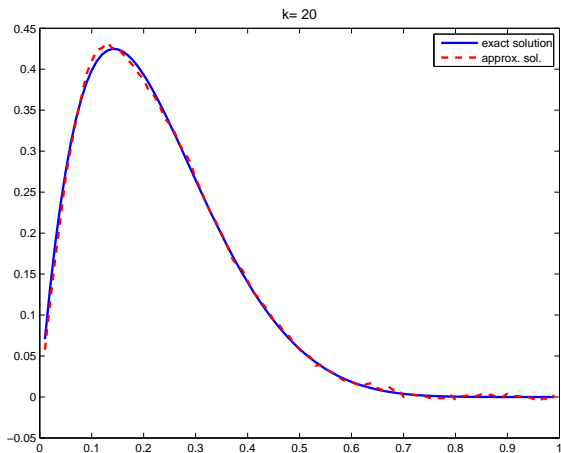
Solutions as function of iteration index



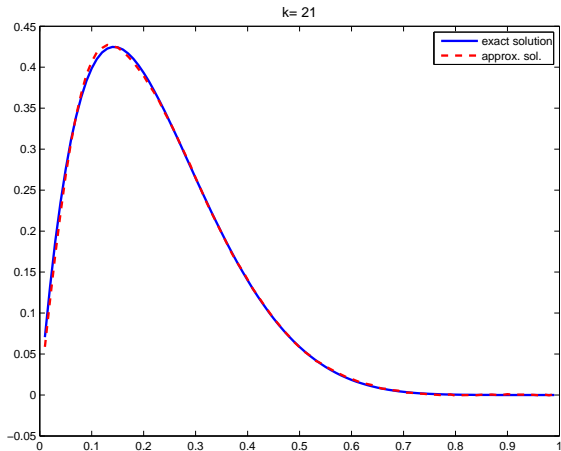
Solutions as function of iteration index



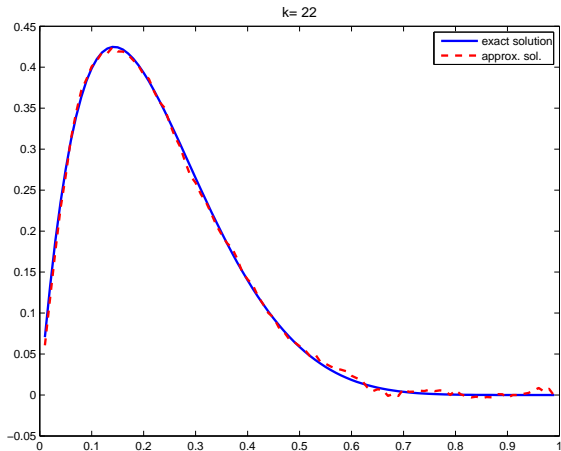
Solutions as function of iteration index



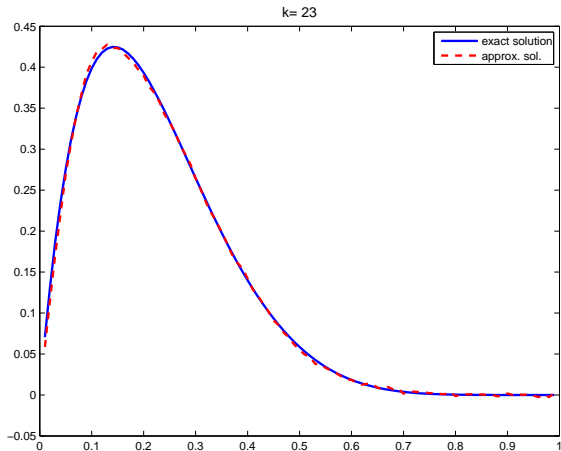
Solutions as function of iteration index



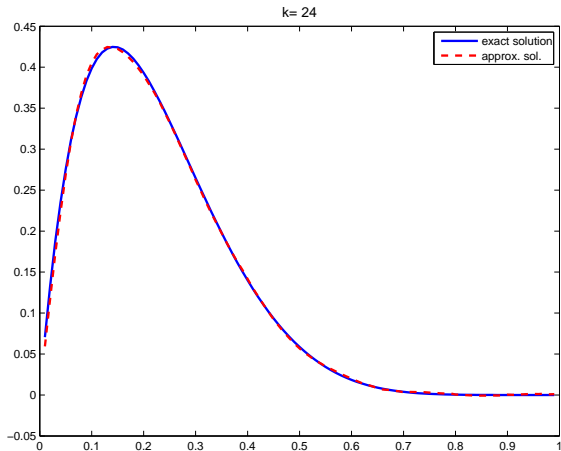
Solutions as function of iteration index



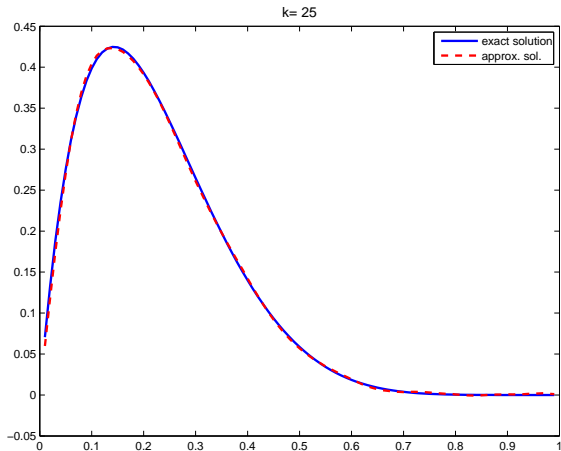
Solutions as function of iteration index



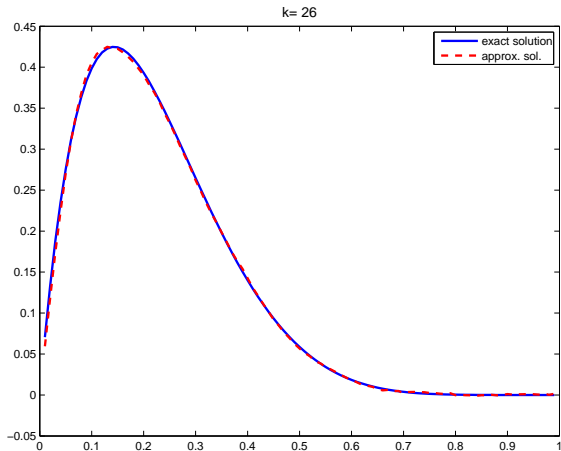
Solutions as function of iteration index



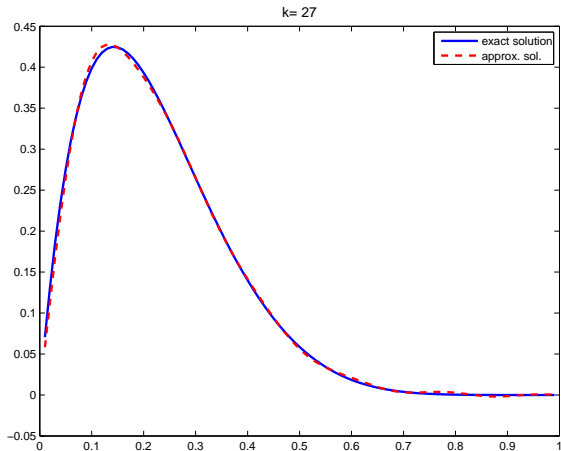
Stopping criterion satisfied here



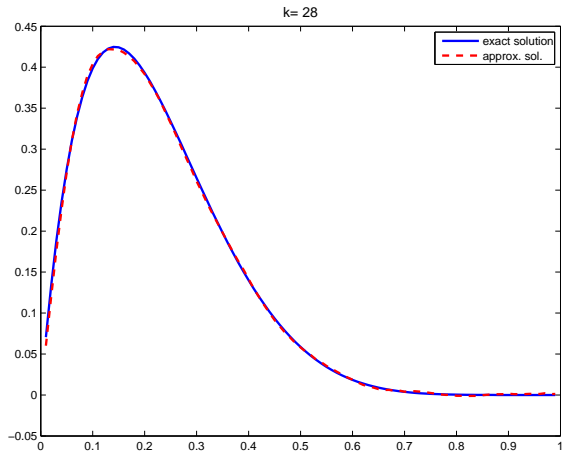
Solutions as function of iteration index



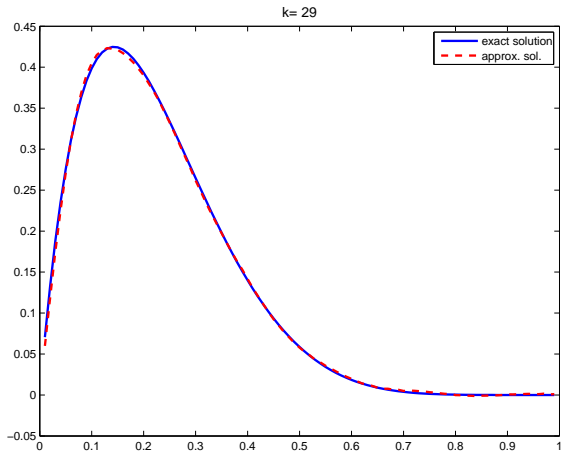
Solutions as function of iteration index



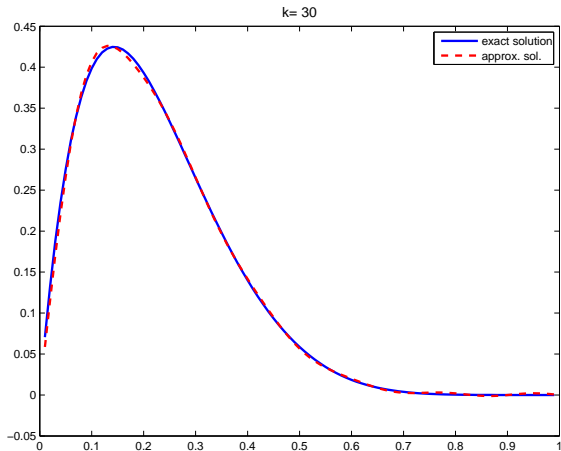
Solutions as function of iteration index



Solutions as function of iteration index



Solutions as function of iteration index



Example 2

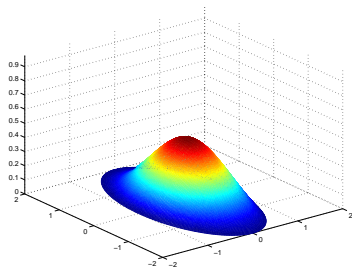
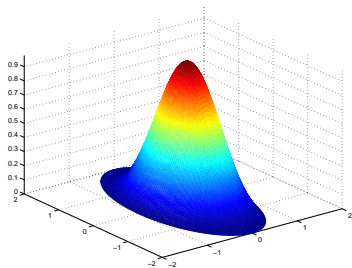
Finite element discretization of L with variable coefficients on the ellipse

$$\Omega = \{(x, y, z) \mid x^2 + y^2/4 \leq 1, 0 \leq z \leq z_1 = 0.6\}.$$

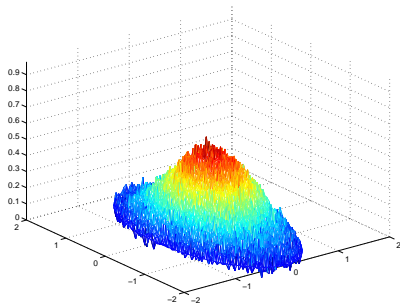
The stiffness matrix has dimension 8065

Data perturbation: $\sim 1\%$

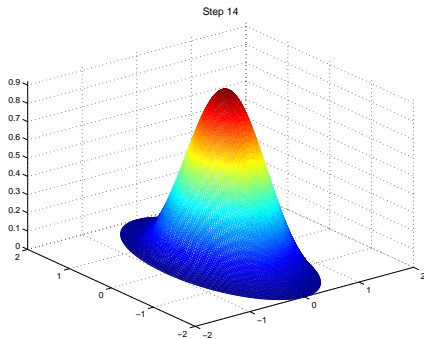
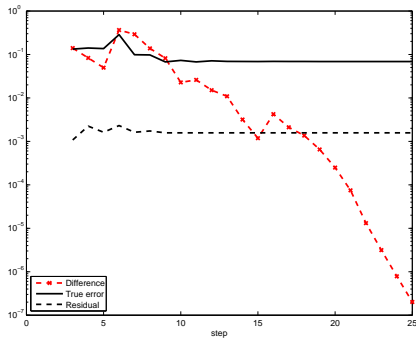
Solution and Exact Data



Perturbed Data



Convergence history ($0.5\lambda_c$). Solution after 14 steps



Conclusions

- 3D Cauchy problem: complex 2D geometry + cylinder in z
- Krylov method + black box 2D elliptic solver
- Stability theory \implies recipe for cut-off level
- Exponential of small matrix is computed (cheap)
- Safe-guarded stopping criterion: Solution difference (cheap) + Residual (rather expensive)
- Much fewer 2D elliptic solves than eigenvalue computation: 98 eigenvalues were smaller than the tolerance.
MATLAB's `eigs`: 300 2D solves
Krylov: 25 solves
 - Highly accurate eigenvalues are not needed
 - The data influence the basis (projection) vectors

Remaining work

- Variable coefficients in z ?
 - Other Cauchy problems (parabolic, Helmholtz, transient electromagnetics)?
-

Talk and paper are available on my web page.