

Preconditioners for Elliptic Control Problems

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Preconditioning KKT Systems

General Formulation of Problem

$$f : \mathbb{R}^n \rightarrow \mathbb{R}, \quad h : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

f, h twice differentiable

$$\text{Minimize } f(x) \quad \text{with } h(x) = 0.$$

Optimal control problem

$$x = (y, u), \quad y \in Y, \quad u \in U$$

System equation (y state, u control, design)

$$h(x) = h(y, u) = y - S(y, u) = 0$$

Lagrangian

$$L(x; l) = f(x) + l^T h(x), \quad x \in \mathbb{R}^n, \quad l \in \mathbb{R}^m,$$

Necessary Optimality Conditions

Theorem

If a constraint qualification holds at optimum x_ , then there exists a Lagrange multiplier λ_* such that*

$$\begin{aligned}L_x(x_*, \lambda_*) &= \nabla f(x_*) - J(x_*)^T \lambda_* &= 0 \\L_\lambda(x_*, \lambda_*) &= h(x_*) &= 0\end{aligned}$$

Sequential **Q**uadratic **P**rogramming similar to Newton's method.
Need to solve **KKT systems**, linear systems of type

$$\begin{pmatrix} L_{xx}(x, \lambda) & J(x) \\ J(x)^T & 0 \end{pmatrix} \begin{pmatrix} \Delta x \\ \Delta \lambda \end{pmatrix} = - \begin{pmatrix} L_x(x, \lambda) \\ h(x) \end{pmatrix}$$

Preconditioning KKT Systems

Splitting of variables: $x = (y, u)$, (state, control or design)

Construction of Preconditioners for linear systems with system matrix

$$K = \begin{pmatrix} L_{yy} & L_{yu} & A^T \\ L_{uy} & L_{uu} & B^T \\ A & B & 0 \end{pmatrix},$$

where

$$\begin{aligned} L_{yy} &\in \mathbb{R}^{m \times m}, & L_{uu} &\in \mathbb{R}^{k \times k}, \\ L_{yu} &\in \mathbb{R}^{m \times k}, & L_{uy} &\in \mathbb{R}^{k \times m}, \\ A &\in \mathbb{R}^{m \times m}, & B &\in \mathbb{R}^{m \times k}. \end{aligned} \quad (k \ll m)$$

An Indefinite Preconditioner for KKT systems Arising in Optimal Control Problems

A. Battermann, EWS 2002

Elliptic Boundary Value Problem

Test Problem

$$\min_{(y,u)} \left\{ \begin{aligned} &\beta_1 \int_{\Gamma_2} Q_2^2(y(x, \bar{z}), u(x)) dx \\ &+ \beta_2 \int_{\Gamma_2} (u(x) - u_v)^2 dx \\ &+ \beta_3 \int_{\Gamma_6} Q_6(y(0, z)) (y_f - y_6) dz \end{aligned} \right\}$$

s.t.

$$\begin{aligned} \Delta y(x, z) &= 0 && \text{in } \Omega, \\ \frac{\partial}{\partial n} y(x, z) &= 0 && \text{on } \Gamma_1 \cup \Gamma_3 \cup \Gamma_5, \\ y(\bar{x}, z) &= y_4 && \text{on } \Gamma_4, \\ y(0, z) &= y_6 && \text{on } \Gamma_6, \\ \frac{\partial}{\partial n} y(x, \bar{z}) &= \frac{1}{d} (u(x) - y(x, \bar{z})) && \text{on } \Gamma_2. \end{aligned}$$

Industrial partner: TGU, Koblenz. (Groundwater modelling)

Elliptic Control Problem

Battermann, Heinkenschloss (1996)

$$\text{Min} \quad \frac{1}{2} \int_{\Omega} (y(x) - y_d(x))^2 dx + \frac{\gamma}{2} \int_{\partial\Omega} u^2(s) ds$$

subject to

$$-\Delta y(x) + y(x) = f(x) \quad x \in \Omega$$

$$\frac{\partial}{\partial n} y(x) = u(x) \quad x \in \partial\Omega$$

and

$$y_{low} \leq y(x) \leq y_{upp} \quad \text{a.e. in } \Omega$$

$$u_{low} \leq u(x) \leq u_{upp} \quad \text{a.e. in } \partial\Omega$$

Structure of KKT Matrix

$$K = \begin{pmatrix} L_{yy} & L_{yu} & A^T \\ L_{uy} & L_{uu} & B^T \\ A & B & 0 \end{pmatrix} = \begin{pmatrix} M_y & 0 & A^T \\ 0 & M_u & B^T \\ A & B & 0 \end{pmatrix}$$

where

- M_y is mass matrix for states
- M_u is mass matrix for controls
- A is stiffness matrix from PDE
- B is (boundary) control input matrix

III-Conditioning

Iterations of GMRES on original system K .
(In all computations, $n_x = 2n_z$ and $k = n$.)

grid size	4	8	16	32
dimension	94	314	1138	4322
#it	8	295	1031	3620

Condition numbers of original system K and submatrix A for different grid sizes.

grid size	4	8	16	32
dimension	94	314	1138	4322
$\kappa(K)$	1356	11238	94713	832836
$\kappa(A)$	48	192	770	3083

Left Preconditioner

$$\tilde{K} = \begin{pmatrix} 0 & 0 & \tilde{A}^T \\ 0 & \tilde{H} & B^T \\ \tilde{A} & B & 0 \end{pmatrix}$$

\tilde{A} preconditioner for A , \tilde{H} preconditioner for

$$H = B^T A^{-T} M_y A^{-1} B + M_u$$

Sometimes $\tilde{H} = M_u$ is sufficient.

\tilde{K} block-triangular, reasonable solves for $\tilde{K}x = -r$

Cost: System and adjoint solve plus solve for H .

Results for GMRES

Iterations of GMRES on preconditioned system $\tilde{K}^{-1}K$.

grid size	4	8	16	32	64	128
dimension	94	314	1138	4322	16834	66434
$\tilde{A} = A, \tilde{H} = H$	3	3	3	3	3	3
$\tilde{A} = A, \tilde{H} = M_U$	5	4	4	4	4	4
$\tilde{A} = A, \tilde{H} = I$	4	5	5	4	4	4
$\tilde{A} = ILU(10^{-4}), \tilde{H} = H$	3	4	4	6	10	16
$\tilde{A} = ILU(10^{-3}), \tilde{H} = H$	4	5	7	11	20	37
$\tilde{A} = ILU(10^{-4}), \tilde{H} = I$	5	7	9	10	15	24
$\tilde{A} = ILU(10^{-3}), \tilde{H} = I$	5	9	13	17	28	50

Convergence Analysis

Theorem (Saad and Schultz)

For the ideal case, i.e. $\tilde{A} = A$, $\tilde{H} = H$, the minimal polynomial of $\tilde{K}^{-1}K$ has degree 3; hence GMRES terminates after 3 steps.

Convergence analysis in non-ideal case difficult.

Therefore consider preconditioners which preserve symmetry.

Block Preconditioners for KKT Systems in PDE-Governed Optimal Control Problems

Battermann, EWS 2003

Preconditioner 1

$$\tilde{K}_1^{-1} = \begin{pmatrix} M_y^{-1/2} & 0 & 0 \\ 0 & M_u^{-1/2} & 0 \\ 0 & 0 & M_y^{1/2} \tilde{A}^{-1} \end{pmatrix}$$

Cost: System and adjoint solve.

This yields the iteration matrix

$$\tilde{K}_1^{-1} K \tilde{K}_1^{-T} = \begin{pmatrix} I & 0 & M_y^{-1/2} A^T \tilde{A}^{-T} M_y^{1/2} \\ 0 & I & M_u^{-1/2} B^T \tilde{A}^{-T} M_y^{1/2} \\ M_y^{1/2} \tilde{A}^{-1} A M_y^{-1/2} & M_y^{1/2} \tilde{A}^{-1} B M_u^{-1/2} & 0 \end{pmatrix}$$

and in the ideal case $\tilde{A} = A$

$$\begin{pmatrix} I & 0 & I \\ 0 & I & M_u^{-1/2} B^T A^{-T} M_y^{1/2} \\ I & M_y^{1/2} A^{-1/2} B M_u^{-1/2} & 0 \end{pmatrix}$$

Analysis of Preconditioner 1

$$K_1 = \tilde{K}_1^{-1} K \tilde{K}_1^{-T} = \begin{pmatrix} I & 0 & I \\ 0 & I & G^T \\ I & G & 0 \end{pmatrix}$$

with $G = M_v^{1/2} A^{-1} B M_u^{-1/2}$.

Note that $I + G^T G$ has same eigenvalues as $M_u^{-1} H$.

Theorem (Battermann, EWS)

Denote the k eigenvalues of $G^T G$ by λ_j . The eigenvalues μ_i of K_1 are

$$\begin{array}{ll} \mu_i = 1 & i = 1, \dots, k \\ \mu_i = 1/2(1 \pm \sqrt{5}) & i = k + 1, \dots, 2m - k \\ \mu_i = 1/2(1 \pm \sqrt{5 + 4\lambda_i}) & i = 2m - k + 1, \dots, 2m + k \end{array}$$

Since $\lambda_i \in [O(h^p), O(1)]$, mesh independence of eigenvalues.

Preconditioner 2

$$\tilde{K}_2^{-1} = \begin{pmatrix} M_y^{-1/2} & 0 & 0 \\ 0 & M_u^{-1/2} & 0 \\ -M_y^{-1/2} & -M_y^{1/2} \tilde{A}^{-1} B M_u^{-1} & M_y^{1/2} \tilde{A}^{-1} \end{pmatrix}$$

Cost: System and adjoint solve.

This yields iteration matrix

$$\tilde{K}_2^{-1} K \tilde{K}_2^{-T} = \begin{pmatrix} I & 0 & -I + C^T \\ 0 & I & 0 \\ -I + C & 0 & I - C - C^T - G G^T \end{pmatrix}$$

with

$$\begin{aligned} G &= M_y^{1/2} \tilde{A}^{-1} B M_u^{-1/2} \\ C &= M_y^{1/2} \tilde{A}^{-1} A M_y^{-1/2} \end{aligned}$$

Analysis Preconditioner 2

In ideal Case $\tilde{A} = A$ implies $C_1 = C_2 = I$

$$K_2 = \tilde{K}_2^{-1} K \tilde{K}_2^{-T} = \begin{pmatrix} I & 0 & 0 \\ 0 & I & 0 \\ 0 & 0 & -I - GG^T \end{pmatrix}$$

Theorem (Battermann, EWS)

Denote the k eigenvalues of GG^T by λ_j . The eigenvalues μ_i of K_2 are given by

$$\begin{aligned} \mu_i &= -1 - \lambda_i & i &= 1, \dots, k \\ \mu_i &= -1 & i &= k + 1, \dots, m \\ \mu_i &= 1 & i &= m + 1, \dots, 2m + k \end{aligned}$$

since $\lambda_i \in [O(h^p), O(1)]$, mesh independence of eigenvalues.

Outlook Preconditioner 2

$$\tilde{K}_2^{-1} K \tilde{K}_2^{-T} = \begin{pmatrix} I & 0 & \Delta^T \\ 0 & I & 0 \\ \Delta & 0 & -I - GG^T - \Delta - \Delta^T \end{pmatrix}$$

with

$$\begin{aligned} G &= M_y^{1/2} A^{-1} B M_u^{-1/2} \\ \Delta &= M_y^{1/2} (\tilde{A}^{-1} A - I) M_y^{-1/2} \end{aligned}$$

Use Bramble-Pasciak preconditioner and apply CG method.
Alternative to CG with constraint preconditioner (projected CG).

Preconditioner 3

Consider ideal case $\tilde{A} = A$

$$\tilde{K}_3^{-1} = \begin{pmatrix} -B^T A^{-T} & I & B^T A^{-T} M_y A^{-1} \\ 0 & 0 & A^{-1} \\ I & 0 & -\frac{1}{2} M_y A^{-1} \end{pmatrix}$$

Cost: 2 system and 2 adjoint solves.

This yields iteration matrix

$$\tilde{K}_3^{-1} K \tilde{K}_3^{-T} = \begin{pmatrix} H & 0 & 0 \\ 0 & 0 & I \\ 0 & I & 0 \end{pmatrix}$$

with $H = M_u + B^T A^{-T} M_y A^{-1} B$.

Additional preconditioning of 1-1-block yields $I + G^T G$ instead of H .

Analysis Preconditioner 3

Ideal Case $\tilde{A} = A$ and preconditioning with M_u .

$$K_3 = \begin{pmatrix} I + G^T G & 0 & 0 \\ 0 & 0 & I \\ 0 & I & 0 \end{pmatrix}$$

Theorem (Battermann, EWS)

Denote the k eigenvalues of GG^T by λ_j . The eigenvalues μ_i of K_3 are given by

$$\begin{aligned} \mu_i &= 1 & i &= 1, \dots, m \\ \mu_i &= -1 & i &= m+1, \dots, 2m \\ \mu_i &= 1 + \lambda_i & i &= 2m+1, \dots, 2m+k \end{aligned}$$

Since $\lambda_i \in [O(h^p), O(1)]$, mesh independence of eigenvalues.

Numerical Comparison (Iteration Count)

GMRES for K and $\tilde{K}^{-1}K$ with $P_A = A, P_H = H$ and $P_A = A, P_H = I$.
 MINRES for precondition. K_1, K_2 and K_3 with $\tilde{A} = A$.

n_z	4	8	16	32	64	128
N	94	314	1138	4322	16834	66434
K	88	294	1028	3666	*	*
$\tilde{K} (P_A = A, P_H = H)$	3	3	3	3	3	3
$\tilde{K} (P_A = A, P_H = I)$	4	4	4	4	4	4
$P_1 (P_A = A)$	18	20	25	30	42	63
$P_2 (P_A = A)$	16	19	22	25	34	47
$P_3 (P_A = A)$	5	6	8	8	8	8

Numerical Comparison (Flop Count)

Computational effort of GMRES and MINRES in megaflops.

n_z	4	8	16	32	64	128
N	94	314	1138	4322	16...	66434
K	2.73	93.44	4K	191K	*	*
$\tilde{K} (P_A = A, P_H = H)$	0.03	0.25	2.90	38.52	548	8,091
$\tilde{K} (P_A = A, P_H = I)$	0.04	0.27	2.61	30.77	418	5,976
$P_1 (P_A = A)$	0.12	0.73	6.41	64.48	779	10,226
$P_2 (P_A = A)$	0.11	0.71	5.94	58.66	708	9,118
$P_3 (P_A = A)$	0.05	0.42	4.33	44.86	528	6,850

Battermann/Heinkenschloss

- Mesh independence observed for all three preconditioners
- Slack variables from control constraints no problem
- Slack variable from state constraints introduce mesh dependence

In the latter case, the matrix

$$G = M_y^{1/2} \tilde{A}^{-1} B M_u^{-1/2}$$

has a large norm due to M_y .

Similar effect for small penalty parameters associated with M_u .

Outlook

- Exploit special structure of preconditioner 2
- Extend to cases with $L_{yu} \neq 0$
- Consider preconditioners \tilde{M}_y, \tilde{M}_u
- Scaling of slack variables
- Influence of penalty parameters
- Partially observed state variables, M_y not invertible