

Generalized Penalty Methods

Solving the resulting subproblems

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Outline

1 Introduction

- Formulation of the (sub-)problem
- Some properties

2 Newton's method with line search

- Formulation of the algorithm
- Convergence analysis

3 Newton's method with smoothed Newton step

- Assumptions and formulation of the algorithm
- Convergence analysis
- Application to model problem

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Optimization problem

In this talk we want to solve problems like

$$\begin{aligned}
 (P_q) \quad & \min_{z \in Z} J_q(z) \\
 \text{subject to} \quad & Ez = 0 \\
 \text{where} \quad & J_q(z) := J(z) + \Psi_q(z) \\
 & := J(z) + \underbrace{\sum_{i=1}^m \int_{\Omega_i} \psi_{i,q_i}(g_i(z)(x) - \varphi_i(x)) dx}_{\Psi_{i,q_i}(z) :=}
 \end{aligned}$$

- Z is a Hilbert space
- $E : Z \rightarrow V$ bounded and linear
 $Z_E := \{z \in Z : Ez = 0\}$
- $g_i : Z \rightarrow L^{r_i}(\Omega_i)$ and $\varphi_i \in L^{r_i}(\Omega_i)$
- $\psi_{i,q} : \mathbb{R} \rightarrow (-\infty, \infty]$ is a penalty function
- Assume q to be fixed

Model problem

$$(MP_q) \quad \min_{z=(y,u)} J_q(z) := J(z) + \Psi_q(z)$$

$$:= \frac{1}{2} \|y - y_d\|_{L^2(\Omega)}^2 + \frac{\gamma}{2} \|u - u_d\|_{L^2(\Omega)}^2 + \Psi_q(z)$$

$$\text{subject to} \quad \begin{aligned} -\Delta y &= u && \text{in } \Omega \\ y &= 0 && \text{on } \partial\Omega \end{aligned}$$

$$\text{where} \quad \Psi_q(z) := \sum_{i=1}^3 \int_{\Omega} \psi_{i,q_i}(g_i(z)(x) - \varphi_i(x)) dx$$

$$g_1(z) := u \quad \varphi_1 := \varphi_u$$

$$g_2(z) := y \quad \varphi_2 := \varphi_y$$

$$g_3(z) := |\nabla y(x)|_2 \quad \varphi_3 := \varphi_g$$

Basic assumptions

- (P1) There is a feasible point z_f
- (P2) The cost functional J is **convex** and lower semi continuous (l.s.c.) on Z_E
- (P3) $\exists \alpha > 0$ s.t. $\forall z_1, z_2 \in Z_E$:

$$J(z_2) \geq J(z_1) + J'(z_1, z_2 - z_1) + \frac{\alpha}{2} \|z_2 - z_1\|^2$$

- (P4) For every closed convex set $U \subset L^r(\Omega_i)$ the pre-image $\{z \in Z_E : g_i(z) \in U\}$ is **closed**. Further all g_i are convex in z (for a.e. $x \in \Omega_i$)
- (Q1) $\psi_{i,q}$ is **convex, l.s.c. and increasing** and $(-\infty, 0) \subset \text{dom} \psi_{i,q}$

Idea

- We have to solve a general non-linear optimization problem with equality constraints
- (Within the model problem) J itself is smooth
- **Idea:** Use a Newton like method
- **But:** Ψ_q is not sufficiently smooth

Additional assumptions on the problem

- (P5a) J is twice continuously differentiable on Z_E
 $D^2J(z)$ is bounded within bounded subsets of Z_E
- (P5b) g_i has the representation $g_i(z)(x) = \eta_i((A_i z)(x))$, where $A_i \in \mathcal{L}(Z, L^2(\Omega_i)^{d_i})$ is a continuous linear operator; $d_i \geq 1$
 $\eta_i : \mathbb{R}^{d_i} \rightarrow \mathbb{R}$ is a convex function

Note that (P5b) does not imply that g_i is twice continuously differentiable, even if η_i is smooth.

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Note that (P5b) does not imply that g_i is twice continuously differentiable, even if η_i is smooth.

Additional assumptions on the problem

(PQ2a) $\pi_i : \mathbb{R}^{d_i} \times \mathbb{R} \rightarrow \mathbb{R}$, $\pi_i(s, t) := \psi_{i,q_i}(\eta_i(s) - t)$ is twice continuously differentiable w.r.t. s for all $t \in \mathbb{R}$
 $\exists L > 0$ const. s.t.

$$|D_{ss}^2 \pi_i(s, t)|_2 \leq L$$

$$|D_{ss}^2 \pi_i(s, t) - D_{ss}^2 \pi_i(\bar{s}, t)|_2 \leq L|s - \bar{s}|_2$$

for all $(s, \bar{s}, t) \in \mathbb{R}^{d_i} \times \mathbb{R}^{d_i} \times \mathbb{R}$ where

$$|D_{ss}^2 \pi(s, t)|_2 := \sup\{\langle D_{ss}^2 \pi_i(s, t)h, k \rangle : |h|_2 = 1, |k|_2 = 1\}$$

(PQ2b) $\Psi_{i,q}$ is continuous on Z

Note that in this setting $\Psi_{i,q}(z) = \int_{\Omega_i} \pi_i((A_i z)(x), \varphi_i(x)) dx$

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(PQ2b) $\Psi_{i,q}$ is continuous on Z

Note that in this setting $\Psi_{i,q}(z) = \int_{\Omega_i} \pi_i((A_i z)(x), \varphi_i(x)) dx$

Discussion for the model problem

The conditions (P5) and (PQ2) are

- not fulfilled for combined logarithmic-quadratic penalty function for original setting and model problem
- **Replace** constraint 3 by:

$$\hat{g}_3(z) := \sqrt{1 + |\nabla y|_2^2} \leq \sqrt{1 + \varphi_g^2} =: \hat{\varphi}_g$$

Then: The conditions are fulfilled for combined penalty function with

$$\begin{aligned} \pi_1(s, t) &= \psi_{1,q}(s - t) & A_1(y, u) &= u \\ \pi_2(s, t) &= \psi_{2,q}(s - t) & A_2(y, u) &= y \\ \pi_3((s_1, s_2), t) &= \psi_{3,q}(\sqrt{1 + |(s_1, s_2)|_2^2} - t) & A_3(y, u) &= \nabla y \end{aligned}$$

Differentiability

Lemma (Gfrerer (5.1) - Differentiability)

Assume: (P4), (P5b), (PQ2)

Then: $\Psi_{i,q}$ is twice Gâteaux-differentiable and $\Psi_{i,q} \in C^{1,1}(Z)$ where

$$\langle D\Psi_{i,q}(z), h \rangle = \int_{\Omega_i} \langle D_s \pi_i(A_i z(x), \varphi_i(x)), A_i h(x) \rangle dx$$

$$\langle D^2 \Psi_{i,q}(z) h, k \rangle = \int_{\Omega_i} \langle D_{ss}^2 \pi_i(A_i z(x), \varphi_i(x)) A_i h(x), A_i k(x) \rangle dx$$

and some continuity result on the second derivative holds:
for all $z \in Z$:

$$\lim_{z' \rightarrow z} \sup_{h \in B, \tilde{k} \in \tilde{K}} \int_{\Omega_i} |\langle D_{ss}^2 \pi_i(A_i z'(x), \varphi(x)) - D_{ss}^2 \pi_i(A_i z(x), \varphi(x)) \rangle h(x), \tilde{k}(x) \rangle| dx = 0$$

Differentiability

Lemma (Gfrerer (5.1) - Differentiability)

Assume: (P4), (P5b), (PQ2)

Then: $\Psi_{i,q}$ is twice Gâteaux-differentiable and $\Psi_{i,q} \in C^{1,1}(Z)$ and some continuity result on the second derivative holds:
for all $z \in Z$:

$$\lim_{z' \rightarrow z} \sup_{h \in B, \tilde{k} \in \tilde{\mathcal{K}}} \int_{\Omega_i} |\langle D_{ss}^2 \pi_i(A_i z'(x), \varphi(x)) - D_{ss}^2 \pi_i(A_i z(x), \varphi(x)) \rangle h(x), \tilde{k}(x) \rangle| dx = 0$$

for every bounded subset $B \subset L^2(\Omega_i)^{d_i}$ and every $\tilde{\mathcal{K}}$ with **either**:

- $\tilde{\mathcal{K}}$ is bounded in $L^{\hat{r}_i}(\Omega_i)^{d_i}$ with $\hat{r}_i > 2$ **or**
- $\tilde{\mathcal{K}} = \{R(x)k(x) : k(x) \text{ belongs to } \mathcal{K} \subset L^2(\Omega_i)^{d_i} \text{ compact and } R \in \mathcal{R} \subset L^\infty(\Omega_i)^{d_i \times d_i} \text{ bounded}\}.$

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Newton's method with line search

If we have a twice differentiable convex function, it is reasonable to apply Newtons' method:

- 1 Choose $0 < \gamma < 1$, $z^0 \in Z_E$; Set $n := 0$
- 2 Compute $h^n \in Z_E$ such that it minimizes

$$\frac{1}{2} \langle D^2 J_q(z^n) h, h \rangle + \langle DJ_q(z^n), h \rangle$$

- 3 Line search: Choose $\sigma_n \in \{1, \frac{1}{2}, \frac{1}{4}, \dots\}$ s.t.

$$J_q(z^n + \sigma_n h^n) \leq J_q(z^n) + \gamma \sigma_n \langle DJ_q(z^n), h^n \rangle$$

- 4 Set $z^{n+1} := z^n + \sigma_n h^n$; Set $n := n + 1$ and goto 2 if stop. crit. is not fulfilled

Newton's method with line search

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Some remarks

- In (MP) the PDE is part of the constraints of the quadratic subproblems.
- The quadratic subproblems

$$\min_{h \in Z_E} \frac{1}{2} \langle D^2 J_q(z^n) h, h \rangle + \langle DJ_q(z^n), h \rangle$$

can be solved e.g. using the optimality system:

Find a stationary point $(h, p) \in Z \times V$ of

$$\frac{1}{2} \langle D^2 J_q(z^n) h, h \rangle + \langle DJ_q(z^n), h \rangle + \langle Eh, p \rangle$$

Leads to KKT-system:

$$\begin{array}{rcl} D^2 J_q(z^n) h & + & E^* p = -DJ_q(z^n) \\ Eh & & = 0 \end{array}$$

Convergence of Newton's method

Theorem (Gfrerer (5.3); Convergence)

Assume (P1) – (P5), (Q1) and (PQ2) and let z^n be generated by Newton's method with line search.

Then: $\lim_{n \rightarrow \infty} z^n = \bar{z}_q$.

If Ψ_q (and therefore J_q) is twice continuously differentiable on Z_E , then the algorithm converges q-superlinear.

Continuous Differentiability

Corollary (Gfrerer (5.2) - Continuous Differentiability)

Assume: (P4), (P5b), (PQ2)

and moreover **either**

- A_i is compact from Z_E into $L^2(\Omega_i)^{d_i}$ **or**
- $A_i \in \mathcal{L}(Z_E, L^{\hat{r}_i}(\Omega_i)^{d_i})$ with $r_i > 2$.

Then: $\Psi_{i,q}$ is twice **continuously** differentiable on Z_E

Model problem

- H_0^1 is compactly embedded in $L^2 \Rightarrow A_2(y, u) = y$ is compact
 \Rightarrow Corr. 5.2 $\Psi_{2,q}$ is twice continuously differentiable
- $-\Delta : H_0^1(\Omega) \rightarrow H^{-1}(\Omega)$ is homeomorphism \Rightarrow
 $\nabla \circ (-\Delta)^{-1} \in \mathcal{L}(H^{-1}(\Omega), L^2(\Omega)^d)$. Since L^2 is compactly
embedded in $H^{-1}(\Omega) \Rightarrow \nabla \circ (-\Delta)^{-1}$ is compact on $L^2 \Rightarrow$
 $A_3(y, u) = \nabla y$ is compact on Z_E
 \Rightarrow Corr. 5.2 $\Psi_{3,q}$ is twice continuously differentiable
- **But:** for $A_1(y, u) = u$ the assumptions of Corr. 5.2 are not
fulfilled.
One can show: $\Psi_{1,q}$ is nowhere twice Fréchet differentiable

So we cannot show q-superlinear convergence.

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Smoothed Newton step

- In this section we modify Newton's algorithm such that the method converges q -superlinear also in the case that Ψ_q is not twice continuously differentiable

Additional assumptions

Recall: (P5b) tells us: $g_i(z) = \eta_i(A_i z)$

(P6a) We can split the constraints ($\exists m'$):

- for $m' + 1, \dots, m$ ("good" constraints):
we have $\Psi_{i,q} \in C^2(Z_E)$

Remark (Corr. 5.3) sufficient:

- $A_i \in \mathcal{L}(Z, L^2(\Omega_i)^{d_i})$ is compact **or**
- $A_i \in \mathcal{L}(Z, L^{\tilde{r}_i}(\Omega_i)^{d_i})$ where $\tilde{r}_i > 2$
- for $i = 1, \dots, m'$ (the others):
 $A_i = \gamma_i B + C_i$ with
 $B \in \mathcal{L}(Z, L^2(\tilde{\Omega})^{\tilde{d}})$ is **common** lin. operator
 $C_i \in \mathcal{L}(Z, L^2(\tilde{\Omega})^{\tilde{d}})$ are **compact** lin. operators
 $\gamma_i \in \mathbb{R}$

Notice: all (these) A_i live in the same spaces (e.g. all $d_i = \tilde{d}$)

(P6b) The mapping $\mathcal{H} : Z \rightarrow V \times L^2(\tilde{\Omega})^{\tilde{d}}$, $\mathcal{H}(z) := (Ez, Bz)$ is surjective.

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Notice: all (these) A_i live in the same spaces (e.g. all $d_i = \tilde{d}$)

(P6b) The mapping $\mathcal{H} : Z \rightarrow V \times L^2(\tilde{\Omega})^{\tilde{d}}$, $\mathcal{H}(z) := (Ez, Bz)$ is surjective.

Newton's method with smoothed Newton step

- 1 Choose $0 < \gamma < 1$, $z^0 \in Z_E$; Set $n := 0$
- 2 Compute $\zeta_1 \in Z$ such that it minimizes

$$\begin{aligned} G_{z^n}(\zeta) &:= \frac{1}{2} \langle D^2 J_q(z^n) \zeta, \zeta \rangle + \langle DJ_q(z^n), \zeta \rangle \\ &= \frac{1}{2} \langle D^2 J(z^n) \zeta, \zeta \rangle + \langle DJ(z^n), \zeta \rangle \\ &\quad + \sum_{i=1}^m \left(\frac{1}{2} \langle D^2 \Psi_{i,q}(z^n) \zeta, \zeta \rangle + \langle D\Psi_{i,q}(z^n), \zeta \rangle \right) \end{aligned}$$

subject to $E\zeta = 0$ and $B\zeta = 0$

Derive a multiplier $(v_1^*, \nu^*) \in V^* \times (L^2(\tilde{\Omega})^{\tilde{d}})^*$ such that

$$DG_{z^n}(\zeta_1) + E^* v_1^* + B^* \nu^* = 0$$

- 3 Compute $\zeta_2 \in \mathcal{U}$
- 4 Compute $\zeta_3 \in Z$
- 5 Set $z^{n+1} := z^n + \zeta_1 + \zeta_2 + \zeta_3$; Set $n := n + 1$ and goto 2 if $\|z^n - z^{n-1}\| > \epsilon$

Newton's method with smoothed Newton step

- 1 Choose $0 < \gamma < 1$, $z^0 \in Z_E$; Set $n := 0$
- 2 Compute $\zeta_1 \in Z$
- 3 Compute $\zeta_2 \in \mathcal{U}$ ($\approx (\text{Ker } B)^\perp$) such that it minimizes

$$\begin{aligned}
 T_n(\zeta) &:= \langle E^* v_1^*, \zeta \rangle \\
 &+ \frac{1}{2} \langle D^2 J(z^n)(\zeta_1 + \zeta), \zeta_1 + \zeta \rangle + \langle DJ(z^n), \zeta_1 + \zeta \rangle \\
 &+ \sum_{i=1}^{m'} \int_{\tilde{\Omega}} (\pi_i((A_i(z^n + \zeta_1) + \gamma_i B\zeta)(x), \varphi_i(x)) + \\
 &\quad \langle D_s \pi_i(A_i(z^n + \zeta_1)), \varphi_i(x) \rangle, C_i \zeta(x)) dx \\
 &+ \sum_{i=m'+1}^m \left(\frac{1}{2} \langle D^2 \Psi_{i,q}(z^n)(\zeta_1 + \zeta), \zeta_1 + \zeta \rangle \right. \\
 &\quad \left. + \langle D \Psi_{i,q}(z^n), \zeta_1 + \zeta \rangle \right)
 \end{aligned}$$

4 Compute $\zeta_3 \in Z$

Newton's method with smoothed Newton step

- 1 Choose $0 < \gamma < 1$, $z^0 \in Z_E$; Set $n := 0$
- 2 Compute $\zeta_1 \in Z$
- 3 Compute $\zeta_2 \in \mathcal{U}$
- 4 Compute $\zeta_3 \in Z$ such that it minimizes

$$\begin{aligned}
 & \frac{1}{2} \langle D^2 J(z^n)(\zeta_1 + \zeta_2 + \zeta_3), \zeta_1 + \zeta_2 + \zeta_3 \rangle + \langle DJ(z^n), \zeta_1 + \zeta_2 + \zeta_3 \rangle \\
 & + \sum_{i=1}^{m'} \left(\frac{1}{2} \langle D^2 \Psi_{i,q}(z^n + \zeta_1 + \zeta_2) \zeta_3, \zeta_3 \rangle + \langle D \Psi_{i,q}(z^n + \zeta_1 + \zeta_2), \zeta_3 \rangle \right) \\
 & + \sum_{i=m'+1}^m \left(\frac{1}{2} \langle D^2 \Psi_{i,q}(z^n)(\zeta_1 + \zeta_2 + \zeta_3), \zeta_1 + \zeta_2 + \zeta_3 \rangle \right. \\
 & \quad \left. + \langle D \Psi_{i,q}(z^n), \zeta_1 + \zeta_2 + \zeta_3 \rangle \right)
 \end{aligned}$$

subject to $E(\zeta_2 + \zeta_3) = 0$

- 5 Set $z^{n+1} := z^n + \zeta_1 + \zeta_2 + \zeta_3$; Set $n := n + 1$ and goto 2, if $\| \nabla J(z^n) \| > \epsilon$

Newton's method with smoothed Newton step

- 1 Choose $0 < \gamma < 1$, $z^0 \in Z_E$; Set $n := 0$
- 2 Compute $\zeta_1 \in Z$
- 3 Compute $\zeta_2 \in \mathcal{U}$
- 4 Compute $\zeta_3 \in Z$
- 5 Set $z^{n+1} := z^n + \zeta_1 + \zeta_2 + \zeta_3$; Set $n := n + 1$ and goto 2 if stop. crit. is not fulfilled

Convergence rate

Lemma (Gfrerer (5.4); convergence rate)

Assume (P1) – (P6), (Q1) and (PQ2).

Then:

$$\|\zeta_1\|_Z + \|\zeta_2\|_Z + \|\zeta_3\|_Z + \|\nu^*\|_{L^2(\tilde{\Omega}^{\tilde{d}})} + \|v_3^* - v_1^*\|_{V^*} = \mathcal{O}(\|z^n - \bar{z}_q\|_{Z_E})$$

Convergence rate

Theorem (Gfrerer (5.5); convergence rate)

Assume (P1) – (P6), (Q1) and (PQ2) and moreover

- that there is some bounded set $\tilde{\mathcal{K}} \subset L^2(\tilde{\Omega})^{\tilde{d}}$ which is either
 - bounded in $L^{\hat{r}}(\tilde{\Omega})^{\tilde{d}}$ with $\hat{r} > 2$ or
 - the elements have the form $\tilde{k}(x) = R(x)k(y)$, where k belongs to a compact subset $\mathcal{K} \subset L^2(\tilde{\Omega})^{\tilde{d}}$ and R or from a bounded subset $\mathcal{R} \subset L^\infty(\tilde{\Omega})^{\hat{d} \times \hat{d}}$

such that for the smoothed Newton steps we have

$\text{dist}(B\zeta_3, \|B\zeta_3\| \tilde{\mathcal{K}}) = o(\|z^n - \bar{z}_q\|_Z)$ for all $z^n \in Z_E$ in some neighborhood of \bar{z}_q .

Then: there exists a increasing function $\omega : \mathbb{R}_+ \rightarrow \mathbb{R}_+$ with $\lim_{t \rightarrow 0_+} \omega(t) = 0$ such that

$$\|z^{n+1} - \bar{z}_q\|_Z \leq \omega(\|z^n - \bar{z}_q\|_Z) \|z^n - \bar{z}_q\|_Z$$

(*q-superlinear convergence*)

Model problem

We can apply Newton's method with smoothed Newton step to (MP)

- **Recall:** $\Psi_{2,q}$ and $\Psi_{3,q}$ are twice continuously differentiable on $Z_E \Rightarrow m' = 1$
- $\Psi_{1,q}(y, u) = u$ is not twice continuously differentiable
But a decomposition as in (P6) is possible:
 $B(y, u) := A_1(y, u) = u$, $\gamma_1 := 1$, $C_1(y, u) := 0$ and
 $\mathcal{U} := \{0\} \times L^2(\Omega) (\approx (\text{Ker} B)^\perp)$

Model problem: the algorithm

- 1 Let $z^n = (y^n, u^n)$ be some iterate
- 2 Find $\zeta_1 := (\zeta_{1,y}, \zeta_{1,u})$ such that it minimizes something
subject to

$$E(\zeta_y, \zeta_u) = 0, \text{ i.e., } -\Delta \zeta_y = \zeta_u \text{ in } \Omega \text{ with } \zeta_y = 0 \text{ on } \partial\Omega$$

$$B(\zeta_y, \zeta_u) = 0, \text{ i.e., } \zeta_u = 0 \text{ on } \Omega.$$

Obviously this PDE has one unique solution: $\zeta_y \equiv 0$.

The multiplier $v_1^* \in H_0^1(\Omega)$ is given by the variational problem

$$\int_{\Omega} (\langle \nabla v_1^*, \nabla v \rangle + (y^n - y_d)v + D_s \pi_2(y^n, \varphi_y)v + \langle D_s \pi_3(\nabla y^n, \hat{\varphi}_g), \nabla v \rangle) = 0 \quad \forall v \in H_0^1(\Omega)$$

- 3 Find $\zeta_2 := (\zeta_{2,y}, \zeta_{2,u})$, where $\zeta_{2,y} = 0$ and $\zeta_{2,u} \in L^2(\Omega)$, where for each $x \in \Omega$ the value $\zeta_{2,u}(x)$ minimizes ($\zeta \in \mathbb{R}$)

$$-v_1^*(x)\zeta + \beta((u^n(x) - u_d(x))\zeta + \frac{1}{2}\zeta^2) + \pi_1(u^n(x) + \zeta, \varphi_u)$$

- 4 Find $\zeta_3 := (\zeta_{3,y}, \zeta_{3,u})$ such that it minimizes

Model problem: the algorithm

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- 2 Find $\zeta_1 := (\zeta_{1,y}, \zeta_{1,u})$ such that it minimizes something
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$$B(\zeta_y, \zeta_u) = 0, \text{ i.e., } \zeta_u = 0 \text{ on } \Omega.$$

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The multiplier $v_1^* \in H_0^1(\Omega)$ is given by the variational problem

$$\int_{\Omega} (\langle \nabla v_1^*, \nabla v \rangle + (y^n - y_d)v + D_s \pi_2(y^n, \varphi_y)v + \langle D_s \pi_3(\nabla y^n, \hat{\varphi}_g), \nabla v \rangle) = 0 \quad \forall v \in H_0^1(\Omega)$$

- 3 Find $\zeta_2 := (\zeta_{2,y}, \zeta_{2,u})$, where $\zeta_{2,y} = 0$ and $\zeta_{2,u} \in L^2(\Omega)$, where for each $x \in \Omega$ the value $\zeta_{2,u}(x)$ minimizes ($\zeta \in \mathbb{R}$)

$$-v_1^*(x)\zeta + \beta((u^n(x) - u_d(x))\zeta + \frac{1}{2}\zeta^2) + \pi_1(u^n(x) + \zeta, \varphi_u)$$

- 4 Find $\zeta_3 := (\zeta_{3,y}, \zeta_{3,u})$ such that it minimizes

Model problem: the algorithm

- 1 Let $z^n = (y^n, u^n)$ be some iterate
- 2 The multiplier $v_1^* \in H_0^1(\Omega)$ is given by ...
- 3 Find $\zeta_2 := (\zeta_{2,y}, \zeta_{2,u})$, where $\zeta_{2,y} = 0$ and $\zeta_{2,u} \in L^2(\Omega)$, where for each $x \in \Omega$ the value $\zeta_{2,u}(x)$ minimizes ($\zeta \in \mathbb{R}$)

$$-v_1^*(x)\zeta + \beta((u^n(x) - u_d(x))\zeta + \frac{1}{2}\zeta^2) + \pi_1(u^n(x) + \zeta, \varphi_u)$$

- 4 Find $\zeta_3 := (\zeta_{3,y}, \zeta_{3,u})$ such that it minimizes

$$\begin{aligned} & \int_{\Omega} (y^n - y_d)\zeta_y + \frac{1}{2}\zeta_y^2 \\ & + \beta((u^n - u_d)(\zeta_u + \zeta_{2,u}) + \frac{1}{2}(\zeta_u + \zeta_{2,u})^2) \\ & + D_s \pi_1(u^n + \zeta_{2,u}, \varphi_u)\zeta_u + \frac{1}{2} D_{ss}^2 \pi_1(u^n + \zeta_{2,u}, \varphi_u)\zeta_u^2 \\ & + D_s \pi_2(y^n, \varphi_y)\zeta_y + \frac{1}{2} D_{ss}^2 \pi_2(y^n, \varphi_y)\zeta_y^2 \\ & + \langle D_s \pi_3(\nabla y^n, \hat{\varphi}_g) + \frac{1}{2} \langle D_{ss}^2 \pi_3(\nabla y^n, \hat{\varphi}_g) \nabla \zeta_y, \zeta_y \rangle \rangle \end{aligned}$$

Model problem: the algorithm

- 1 Let $z^n = (y^n, u^n)$ be some iterate
- 2 The multiplier $v_1^* \in H_0^1(\Omega)$ is given by ...
- 3 Find $\zeta_2 := (\zeta_{2,y}, \zeta_{2,u})$
- 4 Find $\zeta_3 := (\zeta_{3,y}, \zeta_{3,u})$ such that it minimizes

$$\begin{aligned}
 & \int_{\Omega} (y^n - y_d) \zeta_y + \frac{1}{2} \zeta_y^2 \\
 & + \beta((u^n - u_d)(\zeta_u + \zeta_{2,u}) + \frac{1}{2}(\zeta_u + \zeta_{2,u})^2) \\
 & + D_s \pi_1(u^n + \zeta_{2,u}, \varphi_u) \zeta_u + \frac{1}{2} D_{ss}^2 \pi_1(u^n + \zeta_{2,u}, \varphi_u) \zeta_u^2 \\
 & + D_s \pi_2(y^n, \varphi_y) \zeta_y + \frac{1}{2} D_{ss}^2 \pi_2(y^n, \varphi_y) \zeta_y^2 \\
 & + \langle D_s \pi_3(\nabla y^n, \hat{\varphi}_g) + \frac{1}{2} \langle D_{ss}^2 \pi_3(\nabla y^n, \hat{\varphi}_g) \nabla \zeta_y, \zeta_y \rangle \rangle
 \end{aligned}$$

Remarks

The multiplier $v_3^* \in H_0^1(\Omega)$ in step 3 fulfills

$$\begin{aligned} 0 &= \int_{\Omega} (\langle \nabla v_3^*, \nabla v \rangle + (y^n - y_d + \zeta_{3,y} + D_s \pi_2(y^n, \varphi_y) \\ &\quad + D_{ss}^2 \pi_2(y^n, \varphi_y) \zeta_{3,y}) v + \langle D_s \pi_3(\nabla y^n, \hat{\varphi}_g) \\ &\quad + D_{ss}^2 \pi_3(\nabla y^n, \hat{\varphi}_g) \nabla \zeta_{3,y}, \nabla v \rangle) \quad \forall v \in H_0^1(\Omega) \\ 0 &= \beta(u^n - u_d + \zeta_{2,u} + \zeta_{3,u}) + D_s \pi_1(u^n + \zeta_{2,u}, \varphi_u) \\ &\quad + D_{ss}^2 \pi_1(u^n + \zeta_{2,u}, \varphi_u) \zeta_{3,u} - v_3^* \end{aligned}$$

Deduce (using construction of ζ_2) for all $x \in \Omega$

$$(\beta + D_{ss}^2 \pi_1(u^n(x) + \zeta_{2,u}(x), \varphi_u(x))) \zeta_{3,u}(x) = v_3^*(x) - v_1^*(x)$$

By convexity of $\pi_1(s, t)$ w.r.t. s have

$D_{ss}^2 \pi_1(u^n(x) + \zeta_{2,u}(x), \varphi_u(x)) \geq 0$ and since $H_0^1(\Omega)$ is compactly embedded in $L^2(\Omega)$, together with Lemma 5.4, the assumptions of the convergence theorem are fulfilled.

\Rightarrow **algorithm converges for (MP) superlinearly.**

Remarks and Conclusions

- To ensure global convergence:
E.g.: Apply alternating: smoothed Newton step and Newton step with line search
Accept smoothed Newton step only if decrease in objective is achieved
- Numerical results show good results (if the approximation close enough to the exact solution)

Thanks for your attention!

Literature

- H. Gfrerer: Generalized Penalty Methods for a Class of Convex Optimization Problems with Pointwise Inequality Constraints