PDE Constrained Optimization Utilizing Reduced-Order Modeling

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PDE Constrained Optimization Utilizing Reduced-Order Modeling

- Quadratic Programming (QP) Problems
- Nonlinear PDE Constrained Optimization

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Quadratic Programming (QP) Problems

Topics:

- Linear-quadratic optimal control problems
- POD-Galerkin schemes for first-order optimality system
- A-priori and a-posteriori error analysis
- Basis updates
- Regularized state constraints

Linear-Quadratic, Time-Variant Optimal Control Problem

Quadratic programming (QP) problem:

$$\min_{x=(y,u)} J(x) = \frac{1}{2} \|y(t_f) - y_d\|_H^2 + \frac{\kappa}{2} \int_{t_o}^{t_f} \|u(t)\|_U^2 dt$$

subject to the linear evolution problem

$$\frac{\mathrm{d}}{\mathrm{d}t}\langle y(t), \varphi \rangle_H + a(t; y(t), \varphi) = \langle (f + \mathscr{B}u)(t), \varphi \rangle_{V', V} \quad \forall \varphi \in V \text{ a.e. in } (t_\circ, t_f]$$

with $y(t_{\circ}) = y_{\circ}$ in H and to bilateral control constraints

 $u \in \mathscr{U}_{ad} = \{ \tilde{u} \in \mathscr{U} \mid u_a(t) \leq \tilde{u}(t) \leq u_b(t) \text{ a.e. in } [t_\circ, t_f] \}$

- State: $y(t) \in V \hookrightarrow H$ with Hilbert spaces V, H
- Control (Hilbert) space: $\mathscr{U} = L^2(t_\circ, t_f; U)$ with $U = \mathbb{R}^{N_U}$, $U = L^2(\Omega)$ or $U = L^2(\Gamma)$
- Input/control: $u \in \mathscr{U}_{ad}$ (boundary or distributed control)
- Bilinear form: $a(t; \cdot, \cdot)$ continuous and $a(t; \varphi, \varphi) \ge \gamma_1 \|\varphi\|_V^2 \gamma_2 \|\varphi\|_H^2$
- Control operator: $\mathscr{B}: \mathscr{U} \to L^2(t_\circ, t_f; V')$ linear, bounded
- Applicable also for elliptic control problems (Kahlbacher/V:12, Tonn/Urban/V:11, Tröltzsch/V:09)

First-Order Necessary and Sufficient Optimality Conditions

Quadratic programming (QP) problem:

$$\begin{split} \min_{x=(y,u)} J(x) &= \frac{1}{2} \left\| y(t_f) - y_d \right\|_{H}^2 + \frac{\kappa}{2} \int_{t_o}^{t_f} \left\| u(t) \right\|_{U}^2 dt \\ \text{s.t.} \quad \frac{d}{dt} \langle y(t), \varphi \rangle_{H} + a(t; y(t), \varphi) = \langle (t + \mathscr{B}u)(t), \varphi \rangle_{V',V} \ \forall \varphi \in V \text{ a.e. in } (t_o, t_f] \\ y(t_o) &= y_o \text{ in } H \quad \text{and} \quad u_a(t) \leq u(t) \leq u_b(t), \ t \in [t_o, t_f] \end{split}$$

- Optimal state $\overline{\mathbf{y}}$, optimal control $\overline{\mathbf{u}} \in \mathscr{U}_{ad} = \{u | u_a \le u \le u_b \text{ in } [t_\circ, t_f]\}$
- Adjoint/dual equation:

$$-\frac{\mathrm{d}}{\mathrm{d}t}\langle \bar{p}(t), \varphi \rangle_{H} + a(t; \varphi, \bar{p}(t)) = 0 \; \forall \varphi \in V \text{ a.e. in } [t_{\circ}, t_{f}), \quad \bar{p}(t_{f}) = \bar{y}(t_{f}) - y_{d}$$

Variational inequality:

$$\int_{t_{o}}^{t_{f}} \left\langle \kappa \bar{u}(t) - (\mathscr{B}^{\star} \bar{p})(t), u(t) - \bar{u}(t) \right\rangle_{U} \mathrm{d}t \ge 0 \quad \forall u \in \mathscr{U}_{ad}$$
(VI)

• Reduced cost: $\hat{J}(u) = J(\hat{y} + \mathscr{S}u, u)$ with $\hat{J}'(\bar{u}) = \kappa \bar{u} - \mathscr{B}^* \bar{p} \in \mathscr{U}$, i.e.,

$$\langle \hat{J}'(ar{u}), u(t) - ar{u}(t)
angle_{\mathscr{U}} \geq 0 \quad \forall u \in \mathscr{U}_{ad}$$

POD Galerkin for the State Variable

• Particular solution: $\hat{y} \in W(t_{\circ}, t_{f})$ solves $\hat{y}(t_{\circ}) = y_{\circ}$ in H and

$$\frac{\mathrm{d}}{\mathrm{d}t}\langle \hat{\boldsymbol{y}}(t), \boldsymbol{\varphi} \rangle_{H} + \boldsymbol{a}(t; \hat{\boldsymbol{y}}(t), \boldsymbol{\varphi}) = \langle f(t), \boldsymbol{\varphi} \rangle_{V', V} \quad \forall \boldsymbol{\varphi} \in V \text{ a.e. in } (t_{\circ}, t_{f}]$$

- POD space: $V^{\ell} = \operatorname{span} \{ \psi_1, \dots, \psi_{\ell} \} \subset V$
- POD Control-to-state mapping: $\mathscr{S}^{\ell}: \mathscr{U} \to W(t_{\circ}, t_{f}), w^{\ell} = \mathscr{S}^{\ell} u$ solves $w^{\ell}(t_{\circ}) = 0$ in H and

$$\frac{\mathrm{d}}{\mathrm{d}t}\langle \mathsf{w}^{\ell}(t),\psi\rangle_{H} + a(t;\mathsf{w}^{\ell}(t),\psi) = \langle (\mathscr{B}\mathsf{u})(t),\psi\rangle_{V',V} \quad \forall \psi \in V^{\ell} \text{ a.e. in } (t_{\circ},t_{\mathrm{f}}]$$

 $\Rightarrow \mathscr{S}^\ell$ linear and bounded

- POD state: $y^{\ell} = \hat{y} + \mathscr{S}^{\ell} u$
- Previous theorem with a-priori results for the state:
 - a) $y^{\ell}(0) = y_{\circ}$ in *H*, i.e., no POD error in the initial condition

b)
$$\|(\mathscr{S} - \mathscr{S}^{\ell})u\|_{W(t_{\circ}, t_{f})}^{2} \leq \sum_{i \in \mathcal{S}} \lambda_{i}(u)$$

c)
$$\|\mathscr{S} - \mathscr{S}^{\ell}\|_{\mathscr{L}(\mathscr{U},V)} \to 0$$
 for $\ell \to \infty$

POD Galerkin for the Dual Variable (Balancing POD)

• Adjoint/dual equation:

 $-\frac{\mathrm{d}}{\mathrm{d}t}\langle \boldsymbol{p}(t), \boldsymbol{\varphi} \rangle_{H} + \boldsymbol{a}(t; \boldsymbol{\varphi}, \boldsymbol{p}(t)) = 0 \; \forall \boldsymbol{\varphi} \in V \text{ a.e. in } [t_{\circ}, t_{f}), \quad \boldsymbol{p}(t_{f}) = \boldsymbol{y}(t_{f}) - \boldsymbol{y}_{d}$

• Terminal condition: $p(t_f) = y(t_f) - y_d = (\hat{y} + \mathscr{S}u)(t_f) - y_d = \hat{y}(t_f) - y_d + (\mathscr{S}u)(t_f)$

• Particular solution:
$$\hat{p} \in W(t_{\circ}, t_{f})$$
 solves

$$-\frac{\mathrm{d}}{\mathrm{d}t}\langle\hat{\rho}(t),\varphi\rangle_{H}+a(t;\varphi,\hat{\rho}(t))=0\;\forall\varphi\in V\;\mathrm{a.e.\;in}\;[t_{\circ},t_{f}),\quad\hat{\rho}(t_{f})=\hat{y}(t_{f})-y_{d}$$

• Dual solution operator: $\mathscr{A} : \mathscr{U} \to W(t_{\circ}, t_{f}), v = \mathscr{A}u$ solves

$$-\frac{\mathrm{d}}{\mathrm{d}t}\langle \mathsf{v}(t),\varphi\rangle_H + a(t;\varphi,\mathsf{v}(t)) = 0 \;\forall \varphi \in V \text{ a.e. in } [t_\circ,t_f), \quad \mathsf{v}(t_f) = (\mathscr{S}u)(t_f)$$

• POD dual solution operator: $\mathscr{A}^{\ell}: \mathscr{U} \to W(t_{\circ}, t_{f}), v^{\ell} = \mathscr{A}^{\ell} u$ solves

$$-\frac{\mathrm{d}}{\mathrm{d}t}\langle \mathsf{v}^\ell(t),\psi\rangle_H + a(t;\psi,\mathsf{v}^\ell(t)) = 0 \; \forall \psi \in V^\ell \; \text{a.e. in} \; [t_\circ,t_f), \quad \mathsf{v}^\ell(t_f) = (\mathscr{S}^\ell u)(t_f) \in V^\ell$$

 \Rightarrow same POD basis for state and adjoint variable

POD A-Priori Analysis for the Dual Variable

• Continuous variant of POD: for $\ell \in \{1, \dots, d\}$ solve

$$\min\sum_{k=1}^{\ell^2} \int_{t_o}^{t_f} \left\| y^k(t) - \sum_{i=1}^{\ell} \langle y^k(t), \psi_i \rangle_X \psi_i \right\|_X^2 \mathrm{d}t \text{ s.t. } \{\psi_i\}_{i=1}^{\ell} \subset X, \ \langle \psi_i, \psi_j \rangle_X = \delta_{ij}, \ 1 \le i, j \le \ell \quad (\mathbf{P}^{\ell}_{\infty})$$

Theorem (Hinze/V.'08, Tröltzsch/V.'09, Gubisch/V.'13)

$$X = V, \ \mathcal{V} = \text{span} \{ y^{k}(t) | t \in [t_{\circ}, t_{f}], \ 1 \le k \le \emptyset \}$$

a) Snapshots $y^{1} = \mathscr{P}u, \ y^{2} = \mathscr{A}u, \ p = \hat{p} + \mathscr{A}u, \ p^{\ell} = \hat{p} + \mathscr{A}^{\ell}u:$

$$\|p - p^{\ell}\|_{W(t_{\circ}, t_{f})}^{2} \le C_{1} \left(\sum_{i > \ell} \lambda_{i} + \|y_{t}^{1} - \mathscr{P}^{\ell}y_{t}^{1}\|_{L^{2}(t_{\circ}, t_{f}; V)}^{2} + \|y_{t}^{2} - \mathscr{P}^{\ell}y_{t}^{2}\|_{L^{2}(t_{\circ}, t_{f}; V)}^{2} \right)$$

b) Snapshots $y^{1} = \mathscr{P}u, \ y^{2} = \mathscr{A}u, \ y^{3} = (\mathscr{P}u)_{t}, \ y^{4} = (\mathscr{A}u)_{t}, \ all \ in \ L^{2}(t_{\circ}, t_{f}; V):$

$$\|p - p^{\ell}\|_{W(t_{\circ}, t_{f})}^{2} \le C_{2} \sum_{i > \ell} \lambda_{i}$$

c) If $\mathscr{P}\tilde{u} = \mathscr{A}\tilde{u} + t_{i}; \ V$ for all $\tilde{u} \in \mathscr{U}$ then $\lim \|\mathscr{A} - \mathscr{A}^{\ell}\|$ we give $p = 0$

c) If $\mathscr{S}\tilde{u}, \mathscr{A}\tilde{u} \in H^{1}(t_{\circ}, t_{f}; V)$ for all $\tilde{u} \in \mathscr{U}$, then $\lim_{\ell \to \infty} \|\mathscr{A} - \mathscr{A}^{\ell}\|_{\mathscr{L}(\mathscr{U}, V)} = 0$ In particular, $\lim_{\ell \to \infty} \|p(\tilde{u}) - p^{\ell}(\tilde{u})\|_{W(t_{\circ}, t_{f})} = 0$ for any $\tilde{u} \in \mathscr{U}$

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POD Approximation of the Variational Inequality

• Variational inequality:

$$\int_{t_{o}}^{t_{f}} \left\langle \kappa \bar{u}(t) - (\mathscr{B}^{*} \bar{p})(t), u(t) - \bar{u}(t) \right\rangle_{U} \mathrm{d}t \ge 0 \quad \forall u \in \mathscr{U}_{ad}$$
(VI)

• Optimal POD solutions: $\bar{u}^{\ell} \in \mathscr{U}_{ad}$, $\bar{y}^{\ell} = \hat{y} + \mathscr{S}^{\ell} \bar{u}^{\ell}$, $\bar{p}^{\ell} = \hat{p} + \mathscr{A}^{\ell} \bar{u}^{\ell}$

POD variational inequality:

$$\int_{t_{o}}^{t_{f}} \langle \kappa \bar{u}^{\ell}(t) - (\mathscr{B}^{\star} \bar{p}^{\ell})(t), u(t) - \bar{u}^{\ell}(t) \rangle_{U} dt \ge 0 \quad \forall u \in \mathscr{U}_{ad}$$
 (VI^ℓ)

• A-priori analysis: choose $u = \bar{u}^{\ell}$ in (VI), $u = \bar{u}$ in (VI^{ℓ}) and add

$$0 \leq \int_{t_{o}}^{t_{f}} \left\langle \kappa(\bar{u} - \bar{u}^{\ell})(t) - (\mathscr{B}^{\star}(\bar{p} - \bar{p}^{\ell}))(t), \bar{u}^{\ell}(t) - \bar{u}(t) \right\rangle_{U} \mathrm{d}t$$
$$= -\kappa \|\bar{u} - \bar{u}^{\ell}\|_{\mathscr{U}}^{2} \underbrace{-\int_{t_{o}}^{t_{f}} \left\langle (\mathscr{B}^{\star}(\bar{p} - \bar{p}^{\ell}))(t), \bar{u}^{\ell}(t) - \bar{u}(t) \right\rangle_{U} \mathrm{d}t}_{\leq C \|(\mathscr{A} - \mathscr{A}^{\ell})\bar{u}\|_{L^{2}(t_{o}, t_{f}; V)} \|\bar{u}^{\ell} - \bar{u}\|_{\mathscr{U}}}$$

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Convergence Result for the POD Suboptimal Control

• Continuous variant of POD: for $\ell \in \{1, ..., d\}$ solve

$$\min \sum_{k=1}^{\ell^2} \int_{t_o}^{t_f} \left\| y^k(t) - \sum_{i=1}^{\ell} \langle y^k(t), \psi_i \rangle_X \psi_i \right\|_X^2 \mathrm{d}t \text{ s.t. } \{\psi_i\}_{i=1}^{\ell} \subset X, \ \langle \psi_i, \psi_j \rangle_X = \delta_{ij}, \ 1 \le i, j \le \ell \quad (\mathbf{P}_{\boldsymbol{\omega}}^{\ell})$$

Theorem (Hinze/V.'08, Tröltzsch/V.'09, Gubisch/V.'13)

 $X = V, \mathcal{V} = \operatorname{span} \{ \gamma^{k}(t) \mid t \in [t_{\circ}, t_{f}], \ 1 \le k \le \wp \}$

a) Snapshots $y^1 = \mathscr{I}\bar{u}, y^2 = \mathscr{A}\bar{u}, y^3 = (\mathscr{I}\bar{u})_t, y^4 = (\mathscr{A}\bar{u})_t, \text{ all in } L^2(t_\circ, t_f; V):$

$$\|\bar{u}-\bar{u}^{\ell}\|_{W(t_{\circ},t_{f})}^{2} \leq C_{2}\sum_{i>\ell}\lambda_{i}$$

b) If $\mathscr{I}\tilde{u}, \mathscr{A}\tilde{u} \in H^1(t_\circ, t_f; V)$ for all $\tilde{u} \in \mathscr{U}$, then $\lim_{\ell \to \infty} \|\bar{u} - \bar{u}^\ell\|_{\mathscr{U}} = 0$

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Perturbation Analysis (Malanowski/Büskens/Maurer'97)

• Variational inequality:

$$\int_{t_{o}}^{t_{f}} \langle \kappa \bar{u}(t) - (\mathscr{B}^{\star} \bar{p})(t), u(t) - \bar{u}(t) \rangle_{U} dt \ge 0 \quad \forall u \in \mathscr{U}_{ad}$$
(VI)

• Misfit in the variational inequality: suboptimal $\bar{u}^{\ell} \in \mathscr{U}_{ad}$

$$\int_{t_{\circ}}^{t_{f}} \big\langle \kappa \bar{u}^{\ell}(t) - (\mathscr{B}^{\star} \tilde{\rho}^{\ell})(t), u(t) - \bar{u}^{\ell}(t) \big\rangle_{U} \mathrm{d}t \not\geq 0 \quad \forall u \in \mathscr{U}_{ad}$$

with $\tilde{p}^\ell = \hat{p} + \mathscr{A} \bar{u}^\ell$

• Perturbation analysis: there exists a perturbation $\zeta^{\ell} \in \mathscr{U}$ satisfying

$$\int_{t_{\circ}}^{T_{f}} \left\langle \kappa \bar{u}^{\ell}(t) - (\mathscr{B}^{\star} \tilde{\rho}^{\ell})(t) + \zeta^{\ell}(t), u(t) - \bar{u}^{\ell}(t) \right\rangle_{U} \mathrm{d}t \geq 0 \quad \forall u \in \mathscr{U}_{ad}$$

• A-posteriori analysis: choose $u = \overline{u}^{\ell}$ in (VI), $u = \overline{u}$ in (\widetilde{VI}^{ℓ}) and add

$$\kappa \|\bar{u}-\bar{u}^\ell\|_{\mathscr{U}}^2 \leq \int_{t_\circ}^{t_f} \langle (\mathscr{B}^\star\mathscr{A}(\bar{u}^\ell-\bar{u}))+\zeta^\ell(t),\bar{u}^\ell(t)-\bar{u}(t)\rangle_U \mathrm{d} t$$

since $\tilde{p}^{\ell} - \bar{p} = \mathscr{A}(\bar{u}^{\ell} - \bar{u})$

• Estimate for the control: $\|\bar{u} - \bar{u}^{\ell}\|_{\mathscr{U}} \leq \frac{1}{\kappa} \|\zeta^{\ell}\|_{\mathscr{U}}$

Convergence Result for the Perturbation

• Estimate for the control: $\|\bar{u} - \bar{u}^{\ell}\|_{\mathscr{U}} \leq \frac{1}{\kappa} \|\zeta^{\ell}\|_{\mathscr{U}}$ and $\tilde{p}^{\ell} = \hat{p} + \mathscr{A}\bar{u}^{\ell}$

$$\bullet \text{ Computation of } \zeta^{\ell}: \zeta^{\ell}(t) = \begin{cases} -\left(\kappa \overline{u}^{\ell}(t) - (\mathscr{B}^{\star} \widetilde{\rho}^{\ell})(t)\right) & \text{if } u_{\alpha}(t) < \overline{u}^{\ell}(t) < u_{b}(t) \\ -\min\left(0, \kappa \overline{u}^{\ell}(t) - (\mathscr{B}^{\star} \widetilde{\rho}^{\ell})(t)\right) & \text{if } \overline{u}^{\ell}(t) = u_{\alpha}(t) \\ -\max\left(0, \kappa \overline{u}^{\ell}(t) - (\mathscr{B}^{\star} \widetilde{\rho}^{\ell})(t)\right) & \text{if } \overline{u}^{\ell}(t) = u_{b}(t) \end{cases}$$

i.e., $\zeta^{\ell} = \zeta^{\ell}(\bar{u}^{\ell}) \Rightarrow$ a-posteriori error analysis for suboptimal \bar{u}^{ℓ}

Theorem (Tröltzsch/V.'09, Gubisch/V.'13)

$$X = V, \ \mathcal{V} = \operatorname{span} \{ y^{k}(t) \mid t \in [t_{\circ}, t_{f}], \ 1 \le k \le \emptyset \}$$

a) Snapshots $y^{1} = \mathscr{F}\overline{u}, \ y^{2} = \mathscr{A}\overline{u}, \ y^{3} = (\mathscr{F}\overline{u})_{t}, \ y^{4} = (\mathscr{A}\overline{u})_{t}, \ \text{all in } L^{2}(t_{\circ}, t_{f}; V):$
$$\|\zeta^{\ell}\|_{\mathscr{U}}^{2} \le C \sum_{i > \ell} \lambda_{i}$$

b) If $\mathscr{I}\tilde{u}, \mathscr{A}\tilde{u} \in H^1(t_\circ, t_f; V)$ for all $\tilde{u} \in \mathscr{U}$, then $\lim_{\ell \to \infty} \|\zeta^{\ell}\|_{\mathscr{U}} = 0$

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Algorithm with POD A-Posteriori Analysis

• Estimate for the control: $\|\bar{u} - \bar{u}^{\ell}\|_{\mathscr{Y}} \leq \frac{1}{\kappa} \|\zeta^{\ell}\|_{\mathscr{Y}}$ and $\tilde{p}^{\ell} = \hat{p} + \mathscr{A}\bar{u}^{\ell}$

• Computation of
$$\zeta^{\ell}$$
: $\zeta^{\ell}(t) = \begin{cases} -\left(\kappa \overline{u}^{\ell}(t) - (\mathscr{B}^{*} \widetilde{p}^{\ell})(t)\right) & \text{if } u_{a}(t) < \overline{u}^{\ell}(t) < u_{b}(t) \\ -\min\left(0, \kappa \overline{u}^{\ell}(t) - (\mathscr{B}^{*} \widetilde{p}^{\ell})(t)\right) & \text{if } \overline{u}^{\ell}(t) = u_{a}(t) \\ -\max\left(0, \kappa \overline{u}^{\ell}(t) - (\mathscr{B}^{*} \widetilde{p}^{\ell})(t)\right) & \text{if } \overline{u}^{\ell}(t) = u_{b}(t) \end{cases}$

Algorithmus 1 (Optimal control with a-posteriori error estimation)

 Choose ℓ_{max} and POD basis {ψ_i}ℓ_{i=1} for the Galerkin approximation of the LQ problem;
 Determine the reduced-order model for the LQ problem;
 Calculate suboptimal control ūℓ ∈ 𝒜_{ad}, e.g., by a semismooth Newton method;
 Compute perturbation ζℓ = ζℓ(ūℓ);
 IF ||ζℓ||_𝒜 / κ > TOL AND ℓ < ℓ_{max} THEN Enlarge ℓ and go back to (2);
 ELSE Stop; ENDIF

- Applicable for balanced-truncation (Vossen/V/12) or reduced-basis method (Tonn/Urban/V/11)
- Error estimation between high- and low-dimensional discretization (Gubisch/Neitzel)

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Numerical Example: Problem Formulation and Optimal Control (Studinger/V/13)

• Consider:

$$\begin{split} \min J(y, u) &= \frac{1}{2} \int_{\Omega} |y(t_{f}) - y^{cf}|^{2} dx + \frac{1}{200} \int_{0}^{t_{f}} \int_{\Gamma} |u|^{2} d\mathbf{x} dt \\ \text{s.t. } y_{f} + 0.1 \Delta y &= 0 \text{ in } \mathcal{Q}, \ \frac{\partial y}{\partial n} + \frac{y}{100} = u \text{ on } \Sigma, \ y(0) = y_{\circ} \text{ in } \Omega = (0, 1)^{2} \\ -0.5 &\leq u \leq 2 \text{ on } \Sigma = (0, t_{f}) \times \Gamma \end{split}$$

Method & Discretization: semismooth Newton & implizit Euler, finite elements



Numerical Example: POD Error Analysis (Studinger/V.'13)



A-posteriori error: $\|\bar{u} - \bar{u}^{\ell}\|_{\mathscr{U}} \leq \frac{1}{\kappa} \|\zeta^{\ell}(\bar{u}^{\ell})\|_{\mathscr{U}} =: \varepsilon_{\text{ape}}$

l	^ε ape	$\ \bar{u}^h - \bar{u}^\ell\ $	$\frac{\varepsilon_{\text{ape}}}{\ \bar{u}-\bar{u}^{\ell}\ }$	^ε ape	$\ \bar{u}^h - \bar{u}^\ell\ $	$\frac{\varepsilon_{\text{ape}}}{\ \bar{u}-\bar{u}^{\ell}\ }$
5	1.3.10-0	9.1 · 10 ⁻¹	1.32	6.5 · 10 ⁻¹	5.6 · 10 ⁻¹	1.16
20	5.9 · 10 ⁻¹	3.2 · 10 ⁻¹	1.84	7.5·10 ⁻³	7.3 · 10 ⁻³	1.03
60	1.4 · 10 ⁻²	1.2·10 ⁻²	1.17	8.3 · 10 ⁻⁵	8.3 · 10 ⁻⁵	1.00
70	1.2 · 10 ⁻²	1.1.10-2	1.10	3.0.10-5	3.0 · 10 ⁻⁵	1.00
90	1.1.10-2	9.7 · 10 ⁻³	1.13	3.7 · 10 ⁻⁶	3.7 · 10 ⁻⁶	1.00

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Optimality-System POD (OSPOD) (Grimm/V.'13, Kunisch/V.'08, V.'11)

- Original problem: $\min \hat{J}(u) = J(\gamma(u), u)$ s.t. $u \in \mathscr{U}_{ad}$
- POD-Galerkin approximation:

$$\min \hat{J}^{\ell}(u) = J(\gamma^{\ell}(u), u) \quad \text{s.t.} \quad u \in \mathscr{U}_{ad} \tag{\hat{\mathbf{P}}^{\ell}}$$

OSPOD-Problem:

$$\min \hat{\mathscr{J}}^{\ell}(u) = J(y^{\ell}(u), u) \quad \text{s.t.} \begin{cases} y = y(u), \quad \psi_i = \psi_i(u) \text{ for } 1 \le i \le \ell \\ \mathscr{R}\psi_i = \int_{t_o}^{t_f} \langle y(t), \psi_i \rangle_X y(t) \, \mathrm{d}t = \lambda_i \psi_i, \ 1 \le i \le \ell \end{cases} \quad (\hat{\mathbf{P}}^{\ell}_{\mathrm{ospod}})$$

 \rightarrow more complex than the original problem

• Numerical realization: operator splitting

- choose an initial control $u^{(0)}$ and a corresponding POD basis $\{\psi_i^{(0)}\}_{i=1}^{\ell}$
- improve the POD basis by applying a few gradient projection steps for $(\hat{\mathbf{P}}_{asnad}^{\ell})$
- compute an approximate solution to ($\hat{\mathbf{P}}^{\ell}$) by Algorithm 1
- Efficient combination with a-posteriori error estimator (Grimm/V.' 13, V.' 13)

Numerical example for A-Posteriori Error with OSPOD Basis Update (Grimm/V:13)

Algorithmus 1 (Optimal control with a-posteriori error estimation)

(1) Choose ℓ_{\max} and POD basis $\{\psi_i\}_{i=1}^{\ell}$ for the Galerkin approximation of the LQ problem; (2) Determine the reduced-order model for the LQ problem;

(3) Calculate suboptimal control $\bar{u}^{\ell} \in \mathscr{U}_{ad}$, e.g., by a semismooth Newton method;

(4) Compute perturbation $\bar{\zeta}^{\ell} = \zeta^{\ell}(\bar{u}^{\ell})$;

(5) IF $\|\overline{\zeta}^{\ell}\|_{\mathscr{U}}/\kappa$ > TOL AND $\ell < \ell_{\max}$ THEN Enlarge ℓ and go back to (2);

ELSE

Stop; ENDIF

• Algorithm: apply a few gradient steps for ($\hat{\mathbf{P}}_{aspad}^{\ell}$) and continue by Algorithm 1

- Stopping criterium: $\|\bar{\zeta}^{\ell}\|_{\mathscr{U}}/\kappa < \text{TOL} = \min(\Delta t, \Delta x^2) = 1e-4, \ell_{\text{max}} = 30$
- No basis update: $\ell = 10 \nearrow 30$, $\|\bar{\zeta}^{30}\|_{\mathscr{U}} / \kappa \approx 1.2e{-2}$, 48 s for Alg. 1
- 1 OSPOD gradient step: $\ell = 10 \nearrow 30$, $\|\overline{\zeta}^{30}\|_{\mathscr{U}}/\kappa \approx 5e-3$, 4.0s (OSPOD) + 47s for Alg. 1
- 2 OSPOD gradient steps: $\ell = 10 \nearrow 13$, $\|\overline{\zeta}^{13}\|_{\mathscr{U}}/\kappa < \text{TOL}$, 6.3 s (OSPOD) + 8.1 s for Alg. 1

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Problem with Mixed Control-State Constraints (Tröltzsch'05)

• (Abstract) linear-quadratic problem:

$$\begin{split} \min_{\boldsymbol{x}=(\boldsymbol{y},\boldsymbol{u})} J(\boldsymbol{x}) &= \frac{1}{2} \|\boldsymbol{y}(t_{f}) - \boldsymbol{y}_{d}\|_{H}^{2} + \frac{\kappa}{2} \int_{t_{o}}^{t_{f}} \|\boldsymbol{u}(t)\|_{U}^{2} dt \\ \text{s.t.} \quad \frac{d}{dt} \langle \boldsymbol{y}(t), \boldsymbol{\varphi} \rangle_{H} + \boldsymbol{a}(t; \boldsymbol{y}(t), \boldsymbol{\varphi}) &= \langle (t + \mathscr{B}\boldsymbol{u})(t), \boldsymbol{\varphi} \rangle_{V',V} \ \forall \boldsymbol{\varphi} \in V \text{ a.e. in } (t_{o}, t_{f}] \\ \boldsymbol{y}(t_{o}) &= \boldsymbol{y}_{o} \text{ in } H \quad \text{and} \quad \boldsymbol{u}_{a}(t) \leq \varepsilon \boldsymbol{u}(t) + (\mathscr{I}\boldsymbol{y})(t) \leq \boldsymbol{u}_{b}(t), \ t \in [t_{o}, t_{f}] \end{split}$$

• Control space:
$$\mathscr{U} = L^2(t_\circ, t_f; U)$$
 with $U = \mathbb{R}^{N_u}$

• Input/control operator:
$$(\mathscr{B}u)(t,x) = \sum_{i=1}^{N_u} u_i(t)\chi_{\Omega_i}(x)$$

- State operator: $(\mathscr{I} \gamma)(t) = (\int_{\Omega_i} \gamma(t, x) dx / |\Omega_i|)_{1 \le i \le N_u}$
- State constraints: $\varepsilon
 ightarrow 0$
- Particular solution: $\hat{y} \in W(t_{\circ}, t_{f})$ solves $\hat{y}(t_{\circ}) = y_{\circ}$ in H and

$$\frac{\mathrm{d}}{\mathrm{d}t}\langle \hat{y}(t), \varphi \rangle_{H} + a(t; \hat{y}(t), \varphi) = \langle f(t), \varphi \rangle_{V', V} \quad \forall \varphi \in V \text{ a.e. in } (t_{\circ}, t_{f}]$$

• Control-to-state mapping: $\mathscr{S} : \mathscr{U} \to W(t_{\circ}, t_{f}), w = \mathscr{S}u$ solves $w(t_{\circ}) = 0$ in H and

$$\frac{\mathrm{d}}{\mathrm{d} t} \langle \mathsf{w}(t), \varphi \rangle_H + a(t; \mathsf{w}(t), \varphi) = \langle (\mathscr{B} \mathsf{u})(t), \varphi \rangle_{V', V} \quad \forall \varphi \in V \text{ a.e. in } (t_\circ, t_f]$$

• Inequality constraints: $v_a(t) := (u_a - \mathscr{I}\hat{y})(t) \le (\varepsilon u + \mathscr{I}\mathscr{I}u)(t) \le (u_b - \mathscr{I}\hat{y}) =: v_b(t)$

Formulation as a Control Constrained Problem

- Solution representation: $u \mapsto y(u) = \hat{y} + \mathscr{S}u$
- Inequality constraints: $v_a(t) := (u_a \mathscr{I}\hat{y})(t) \le (\varepsilon u + \mathscr{I}\mathscr{I}u)(t) \le (u_b \mathscr{I}\hat{y}) =: v_b(t)$
- Transformation of variables: $v := \mathscr{F}u$ with $\mathscr{F} = \varepsilon + \mathscr{I}\mathscr{S}$, i.e. $y = \hat{y} + \mathscr{S}u = \hat{y} + \mathscr{I}\mathscr{F}^{-1}v$
- Transformed, linear-quadratic problem:

$$\min_{\mathbf{v}} J(\hat{\mathbf{v}} + \mathscr{SF}^{-1}\mathbf{v}, \mathscr{F}^{-1}\mathbf{v}) = \dots + \frac{\kappa}{2} \|\mathscr{F}^{-1}\mathbf{v}\|^2$$

s.t. $\mathbf{v} \in V_{\text{cd}} = \{\tilde{\mathbf{v}} | v_{\mathbf{q}}(t) \le \tilde{\mathbf{v}}(t) \le v_{\mathbf{b}}(t) \text{ for } t \in [t_{\circ}, t_{f}]\}$

 \Rightarrow form of the previous linear-quadratic optimal control problem

• Variational inequality:

$$\left\langle \kappa \mathscr{F}^{-\star} \mathscr{F}^{-1} \bar{v}(t) - \mathscr{B}^{\star} \bar{\rho}(t), v(t) - \bar{v}(t) \right\rangle_{\mathscr{U}} \geq 0 \quad \forall v \in V_{ad}$$

- POD Galerkin scheme: additional analysis for $\mathscr{F}^{\ell} \approx \mathscr{F}$
- A-posteriori error estimate: $\|\bar{u} \bar{u}^{\ell}\| \leq \frac{1}{\kappa} \|\mathscr{F}\|_{\mathscr{L}(\mathscr{U})} \|\zeta^{\ell}\|$ with computable ζ^{ℓ}
- Convergence result: $\|\zeta^{\ell}\| \to 0$ for $\ell \to \infty$

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Numerical Example: POD basis update (Afanasiev/Hinze'01, Gubisch/V.'13)

• Optimal control problem: heat equation, control space $\mathscr{U} = L^2(0,3;\mathbb{R}^{10})$, $\varepsilon = 1e-5$



Stefan Volkwein

PDE Constrained Optimization Utilizing Reduced-Order Modeling

Related Literature

- Afanasiev, Dede, Fahl, Grepl, Hinze, Kärcher, Manzoni, Negri, Quarteroni, Sachs, Schu...
- M. Gubisch and S. Volkwein: POD a-posteriori error analysis for optimal control problems with mixed control-state constraints, 2013
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Nonlinear PDE Constrained Optimization

Topics:

- Inexact sequential quadratic programming (SQP)
- A-posteriori error control for nonlinear problems

Sequential Quadratic Programming (SQP)

Infinite dimensional optimization:

$$\min J(x) \quad \text{s.t.} \quad e(x) = 0 \tag{P}$$

- Lagrange functional for (P): $\mathscr{L}(x,p) = J(x) + \langle e(x), p \rangle$
- (Local) SQP method: at $z_k = (x_k, p_k)$ solve

$$\begin{cases} \min_{x_{\delta}} \mathscr{L}_{x}(z_{k}) x_{\delta} + \frac{1}{2} \mathscr{L}_{xx}(z_{k})(x_{\delta}, x_{\delta}) \\ \text{s.t. } e(x_{k}) + e'(x_{k}) x_{\delta} = 0 \end{cases}$$

• KKT system: solution \bar{x}_{δ} to (**QP**^k) is characterized by

$$\underbrace{\begin{pmatrix} \mathscr{L}_{xx}(z_k) & e'(x_k)^* \\ e'(x_k) & 0 \end{pmatrix}}_{A_k} \underbrace{\begin{pmatrix} \bar{x}_{\delta} \\ \bar{p}_{\delta} \end{pmatrix}}_{\bar{z}_{\delta}} = \underbrace{-\begin{pmatrix} \mathscr{L}_x(z_k) \\ e(x_k) \end{pmatrix}}_{b_k}$$

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 (\mathbf{QP}^k)

Inexact SQP by Using POD or RB

- **KKT system**: inexact solve of $A_k \bar{z}_{\delta} = b_k$ by discretization
- Discretization: (POD or RB or BT or...) model reduction

$$A_k^\ell \bar{z}_\delta^\ell = b_k^\ell \in \mathbb{R}^n, \quad n = n(\ell)$$

• Convergence of (local) SQP method: \bar{z}^{ℓ}_{δ} reduced-order solution

$$\|A_k \mathscr{P} \overline{z}_{\delta}^{\ell} - b_k\| = \mathscr{O}(\|\mathscr{L}'(z_k)\|^q), \quad q \in [1, 2]$$

with prolongation \mathcal{P}

- Rate of convergence: superlinear (1 < q < 2), quadratic (q = 2)
- Control of reduced-order approach:

$$\|A_k\mathscr{P}\bar{z}_{\delta}^{\ell}-b_k\|\simeq\|\bar{z}_{\delta}-\mathscr{P}\bar{z}_{\delta}^{\ell}\|\simeq\|\mathscr{L}'(z_k)\|^q$$

Convergence Result

• Variables in optimal control: x = (y, u), y = y(u)

• KKT system:
$$z_k = (x_k, p_k), x_k = (y_k, u_k)$$

$$\begin{pmatrix} \mathcal{L}_{YY}(z_k) & \mathcal{L}_{YU}(z_k) & | & e_Y(x_k)^* \\ \mathcal{L}_{UY}(z_k) & \mathcal{L}_{UU}(z_k) & | & e_U(x_k)^* \\ \hline e_Y(x_k) & e_U(x_k) & | & 0 \end{pmatrix} \begin{pmatrix} & y_\delta \\ & U_\delta \\ & p_\delta \end{pmatrix} = \begin{pmatrix} & -\mathcal{L}_Y(z_k) \\ & -\mathcal{L}_U(z_k) \\ & -e(x_k) \end{pmatrix}$$

- Suboptimal solution to KKT system: $\bar{z}_{\delta}^{\ell} = (\bar{y}_{\delta}^{\ell}, \bar{u}_{\delta}^{\ell}, \bar{p}_{\delta}^{\ell})$
- Prolongation $\mathscr{P}: \bar{z}^{\ell}_{\delta} \mapsto \mathscr{P}\bar{z}^{\ell}_{\delta} = (\tilde{y}_{\delta}, \bar{u}^{\ell}_{\delta}, \tilde{\rho}_{\delta})$ with

$$\begin{split} & \Theta_{Y}(x_{k})\tilde{Y}_{\delta} = -\Theta(x_{k}) - \Theta_{U}(x_{k})\tilde{U}_{\delta}^{\ell} \\ & \Theta_{Y}(x_{k})^{*}\tilde{\rho}_{\delta} = -\mathcal{L}_{Y}(z_{k}) - \mathcal{L}_{YY}(z_{k})\tilde{Y} - \mathcal{L}_{YU}(z_{k})\tilde{U}_{\delta}^{\ell} \end{split}$$

 \rightarrow a-posteriori error computable

Theorem (Kahlbacher/V.'12)

Second-order sufficient optimality implies

$$\lim_{k \to \infty} z_k + \mathscr{P} \bar{z}_{\delta}^{\ell} = \bar{z} \quad \text{if} \quad \|A_k \mathscr{P} \bar{z}_{\delta}^{\ell} - b_k\| \simeq \|\bar{u}_{\delta} - \bar{u}_{\delta}^{\ell}\| < \text{TOL}$$

Multilevel Approach with Reduced-Order Models

- Convergence criterium: $||A_k \mathscr{P} \overline{z}_{\delta}^{\ell} b_k|| \simeq ||\overline{u}_{\delta} \overline{u}_{\delta}^{\ell}|| < \text{TOL}$
- A-posteriori error (Tröltzsch/V.'09):

$$\|\bar{u}_{\delta} - \bar{u}_{\delta}^{\ell}\| \simeq \|\underbrace{\mathscr{L}_{\mathcal{U}\mathcal{Y}}(z_k)\tilde{y}_{\delta} + \mathscr{L}_{\mathcal{U}\mathcal{U}}(z_k)\bar{u}_{\delta}^{\ell} + e_u(x_k)^{\star}\tilde{\rho}_{\delta} + \mathscr{L}_{\mathcal{U}}(z_k)}_{:=-\bar{\zeta}^{\ell}}\|$$

with $\| \overline{\zeta}^\ell \| \to 0$ for $\ell \to \infty$

- Convergence of $\|\bar{\zeta}^{\ell}\|$: no rate, basis dependent (Hinze/V.'08)
- POD basis: combination with Optimality-System POD (V(11))
- Nonlinear optimization approach: Trust-Region POD (Arian/Fahl/Sachs'00, Schu/Sachs'07)
- Combination with FE adaptivity: (Clever/Lang/Ulbrich/Ziems)

Numerical experiments (Kahlbacher/V.'12)

Optimal control problem:

$$\begin{split} \min \frac{1}{2} \int_{\Omega_m} |y - y^d|^2 \, dx + \frac{\kappa}{2} \sum_{i=1}^{n_\Omega} |\Omega_i| |u_i - u_i^\circ|^2 \\ \text{s.t. } u_i \ge 0, \quad -\Delta y + y \sum_{i=1}^{n_\Omega} u_i \chi_{\Omega_i} = f \text{ in } \Omega, \ \frac{\partial y}{\partial n} + y = g \text{ on } \Gamma \end{split}$$

- Given data: y^d , u_i° , $\kappa > 0$, f, g
- Globalisation of SQP:
 - modification of the hessian
 - Armijo linesearch with ℓ_1 merit function
- Equality constraint case: $\bar{x} = (\bar{y}, \bar{u})$ with inactive $\bar{u} > 0$

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Numerical examples (Kahlbacher/V.'12)



Reference control: u ^o	
Desired state: $y(u^\circ)$ + noise	
Measurement domain: Ω_m	2



SQP it.	$ \ \mathscr{L}'(\mathbf{z}_k) \ $	a-post.
k = 0 k = 1 k = 2 k = 3 k = 4 k = 5	1.38e-0 8.53e-1 2.57e-1 4.65e-3 3.05e-5 4.66e-9	1.16e-3 8.36e-2 7.87e-5 5.67e-6 1.90e-6

SQP it.	$\ \mathscr{L}'(z_k)\ $	a-post.
k = 0 k = 1 k = 2 k = 3 k = 4	3.01e-2 4.48e-1 3.63e-1 5.16e-2	1.58e-3 7.44e-4 3.46e-4 4.04e-5
k = 4 k = 5 k = 6	1.40e-2 1.62e-3 7.51e-7	4.96e-4 6.56e-4 —

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Elliptic-Parabolic Systems Arising in Battery Models (DEN'96, WXZ'06)

• Coupled nonlinear parameterized PDE system: $\Omega = (0, L)$

$$\begin{aligned} \mathbf{y}_{t} - \nabla \cdot (D\nabla \mathbf{y}) - \mathscr{F}(\mathbf{y}, \mathbf{p}, q; \mu) &= 0 & \text{in } \mathbf{Q} = (t_{\circ}, t_{f}) \times \Omega \\ -\nabla \cdot (\kappa(\mathbf{y}; \mu) \nabla \mathbf{p}) - \mathscr{F}(\mathbf{y}, \mathbf{p}, q; \mu) &= 0 & \text{in } \mathbf{Q} \\ -\nabla \cdot (\sigma \nabla q) + \mathscr{F}(\mathbf{y}, \mathbf{p}, q; \mu) &= 0 & \text{in } \mathbf{Q} \end{aligned}$$

 \rightarrow concentration y and potentials p, q

• Parameterized nonlinearities: $\mu = (\mu_1, \mu_2, \mu_3, \mu_4) \in \mathbb{R}^4$

$$\kappa(\boldsymbol{y};\boldsymbol{\mu}) = (1 + \mu_{4}\boldsymbol{y})^{3}, \quad \mathscr{F}(\boldsymbol{y},\boldsymbol{p},\boldsymbol{q};\boldsymbol{\mu}) = \begin{cases} \begin{array}{c} \mu_{2}\sqrt{y} \mathrm{sinh}(\mu_{1}(\boldsymbol{q}-\boldsymbol{p}-\ln\boldsymbol{y})) & \mathrm{in} \ \Omega_{s}^{1} \\ 0 & \mathrm{in} \ \Omega_{\Theta} \\ \mu_{3}\sqrt{y} \mathrm{sinh}(\mu_{1}(\boldsymbol{q}-\boldsymbol{p}-\ln\boldsymbol{y})) & \mathrm{in} \ \Omega_{s}^{2} \end{cases}$$

• Boundary conditions: $y_x(t, \cdot) = p_x(t, \cdot) = 0$, q(t, 0) = 0 (uniqueness), $\sigma q_x(t, L) = l(t)$



Variational Formulation of the PDEs

- Coupled nonlinear parameterized PDE system: $\Omega = (0, L)$
 - $\begin{aligned} \mathbf{y}_{t} \nabla \cdot (D\nabla \mathbf{y}) \mathscr{F}(\mathbf{y}, \mathbf{p}, \mathbf{q}; \mu) &= 0 \\ -\nabla \cdot (\kappa(\mathbf{y}; \mu) \nabla \mathbf{p}) \mathscr{F}(\mathbf{y}, \mathbf{p}, \mathbf{q}; \mu) &= 0 \end{aligned} \qquad \text{in } \mathbf{Q} = (t_{\circ}, t_{f}) \times \Omega \\ \mathbf{u}_{t} = \mathbf{u}_{t} + \mathbf{u}$

$$-\nabla \cdot (\sigma \nabla q) + \mathscr{F}(\gamma, p, q; \mu) = 0 \qquad \text{in } Q$$

and boundary conditions

• Variational formulation: $V_0 = \{ \varphi \in V : \varphi(0) = 0 \} \subset V = H^1(\Omega)$

$$\int_{\Omega} \mathbf{y}_{f} \boldsymbol{\varphi} + D \nabla \mathbf{y} \cdot \nabla \boldsymbol{\varphi} - \mathscr{F}(\mathbf{y}, \mathbf{p}, \mathbf{q}; \boldsymbol{\mu}) \boldsymbol{\varphi} \, \mathrm{d} \boldsymbol{x} = \boldsymbol{0} \qquad \qquad \forall \boldsymbol{\varphi} \in \boldsymbol{V}$$

$$\int_{\Omega} \kappa(\boldsymbol{y};\boldsymbol{\mu}) \nabla \boldsymbol{p} \cdot \nabla \boldsymbol{\varphi} - \mathscr{F}(\boldsymbol{y},\boldsymbol{p},\boldsymbol{q};\boldsymbol{\mu}) \boldsymbol{\varphi} \, \mathrm{d} \boldsymbol{x} = \boldsymbol{0} \qquad \qquad \forall \boldsymbol{\varphi} \in \boldsymbol{V}$$

$$\int_{\Omega} \sigma \nabla \boldsymbol{q} \cdot \nabla \boldsymbol{\chi} + \mathscr{F}(\boldsymbol{\gamma}, \boldsymbol{p}, \boldsymbol{q}; \boldsymbol{\mu}) \boldsymbol{\chi} \, \mathrm{d} \boldsymbol{x} = l \boldsymbol{\chi}(L) \qquad \forall \boldsymbol{\chi} \in V_0$$



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PDE Constrained Optimization Utilizing Reduced-Order Modeling

Finite Element (FE) Galerkin Model (Lass/V.'13)

• Nonlinear FE-Galerkin scheme: $\gamma^{\mathscr{N}}(t) = \sum_{i=1}^{\mathscr{N}} \underline{y}_{i}(t) \varphi_{i}^{\mathscr{N}}$ etc.

$$\int_{\Omega} \boldsymbol{\gamma}_{T}^{\mathcal{N}} \boldsymbol{\varphi}^{\mathcal{N}} + D \nabla \boldsymbol{\gamma}^{\mathcal{N}} \cdot \nabla \boldsymbol{\varphi}^{\mathcal{N}} - \mathscr{F}(\boldsymbol{\gamma}^{\mathcal{N}}, \boldsymbol{p}^{\mathcal{N}}, \boldsymbol{q}^{\mathcal{N}}; \boldsymbol{\mu}) \boldsymbol{\varphi}^{\mathcal{N}} \, \mathrm{d} \boldsymbol{x} = 0 \qquad \qquad \forall \boldsymbol{\varphi}^{\mathcal{N}} \in V^{\mathcal{N}}$$

$$\int_{\Omega} \kappa(\boldsymbol{y}^{\mathcal{N}};\boldsymbol{\mu}) \nabla \boldsymbol{p}^{\mathcal{N}} \cdot \nabla \boldsymbol{\varphi}^{\mathcal{N}} - \mathscr{F}(\boldsymbol{y}^{\mathcal{N}},\boldsymbol{p}^{\mathcal{N}},\boldsymbol{q}^{\mathcal{N}};\boldsymbol{\mu}) \boldsymbol{\varphi}^{\mathcal{N}} \, \mathrm{d} \boldsymbol{x} = \boldsymbol{0} \qquad \qquad \forall \boldsymbol{\varphi}^{\mathcal{N}} \in \mathsf{V}^{\mathcal{N}}$$

$$\int_{\Omega} \sigma \nabla \boldsymbol{q}^{\mathcal{N}} \cdot \nabla \boldsymbol{\chi}^{h} + \mathscr{F}(\boldsymbol{y}^{\mathcal{N}}, \boldsymbol{p}^{\mathcal{N}}, \boldsymbol{q}^{\mathcal{N}}; \boldsymbol{\mu}) \boldsymbol{\chi}^{\mathcal{N}} \, \mathrm{d}\boldsymbol{x} = l \boldsymbol{\chi}^{\mathcal{N}}(L) \qquad \forall \boldsymbol{\chi}^{\mathcal{N}} \in V_{0}^{\mathcal{N}}$$

• High-dimensional (large *N*) nonlinear ODE system:

$$\mathbf{M}\underline{\mathbf{y}}'(t) + \mathbf{S}_{D}\underline{\mathbf{y}}(t) - \mathbf{F}(\underline{\mathbf{y}}(t),\underline{\mathbf{p}}(t),\underline{\mathbf{q}}(t);\mu) = \mathbf{0}, \qquad t \in (t_{\circ},t_{f}]$$

$$S_{\kappa}(\underline{\gamma}(t);\mu)\underline{\rho}(t) - F(\underline{\gamma}(t),\underline{\rho}(t),\underline{q}(t);\mu) = 0, \qquad t \in (t_{\circ},t_{f}]$$

$$\mathbf{S}_{\sigma}\underline{\mathbf{q}}(t) + \mathbf{F}(\underline{\mathbf{y}}(t),\underline{\mathbf{p}}(t),\underline{\mathbf{q}}(t);\boldsymbol{\mu}) = \underline{\mathbf{l}}(t), \qquad t \in (t_{\circ},t_{f}]$$

$$\begin{split} M & ... \text{ mass matrix, } S_* & ... \text{ stiffness matrices depending on the diffusion parameters } D, \ \kappa(\gamma;\mu), \sigma \\ F(\underline{z}(f);\mu) &= (\int_{\Omega} \mathscr{F}(\gamma^{\mathcal{A}}(t); p^{\mathcal{A}}(t); q^{\mathcal{A}}(t); \mu) \varphi_j^{\mathcal{A}} dx)_j \text{ with coeffizient vector } \underline{z} = (\underline{\gamma}, \underline{p}, \underline{q}) \end{split}$$

• Model reduction: replace $\{\varphi_i^{\mathcal{N}}\}_{i=1}^{\mathcal{N}}$ by $\{\psi_i\}_{i=1}^{\ell}$ with $\ell \ll \mathcal{N}$

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Parameter Estimation without Noise (Lass'13)

- Optimization problem: estimate $\mu = (\mu_1, \dots, \mu_4)$ from boundary potential data $q^d(t)$
- Cost functional: $\hat{J}(\mu) = \frac{1}{2} \int_{t_{-}}^{t_{f}} |q(t,L;\mu) q^{d}(t)|^{2} dx$
- Subset selection method: determine parameters, which can be identified (Kappel et al. 12)



- Numerical algorithm: Gauss-Newton method (regularisation by subset selection)
- Computation times (Noise: 0%):

FE 358.8 s (0.80000, -0.90000, -0.20000, 0.10000) POD 9.9 s (0.79995, -0.90000, -0.20003, 0.10001)		CPU time	optimal parameters
· · · · · · · · · · · · · · · · · · ·	FE	358.8 s	(0.80000, -0.90000, -0.20000, 0.10000)
	POD	9.9 s	(0.79995, -0.90000, -0.20003, 0.10001)

Parameter Estimation with Noise (Lass' 13)

- Optimization problem: estimate $\mu = (\mu_1, \dots, \mu_4)$ from boundary potential data $q^d(t)$
- Cost functional: $\hat{J}(\mu) = \frac{1}{2} \int_{t}^{t_{f}} |q(t,L;\mu) q^{d}(t)|^{2} dx$
- Subset selection method: determine parameters, which can be identified (Kappel et al. 12)



- Numerical algorithm: Gauss-Newton method (regularisation by subset selection)
- Computation times (Noise: 5%):

	FOIIIIE	oplittidi parameters
FE	314.6s	(0.77324, -0.90000, -0.19157, 0.15786)
POD	9.3s	(0.77310, -0.90000, -0.19156, 0.15806)

A-Posteriori Analysis for Nonlinear Optimal Control Problems (Kammann/Tröltzsch/V.'12, DFG grant)

- **Reduced problem**: $\min_{\mu \in \mathfrak{D}} \hat{J}(\mu)$ with hessian $\nabla^2 \hat{J}(\mu)$
- First-order optimality conditions: $\nabla \hat{J}(\bar{\mu})^{\top}(\mu \bar{\mu}) \ge 0$ for all $\mu \in \mathfrak{D}$
- Second-order sufficient optimality conditions: there is a constant $\eta = \eta(\bar{\mu}) > 0$ with

 $\mu^{ op} \hat{J}''(\bar{\mu})\mu \geq \eta \|\mu\|_2^2 \quad \forall \mu \in \mathbb{R}^4$

 $\Rightarrow \mu^{\top} \nabla^2 \hat{J}(\tilde{\mu}) \mu \geq \frac{\eta}{2} \|\mu\|_2^2 \ \forall \mu \in \mathbb{R}^4, \forall \tilde{\mu} \in \mathfrak{D}_{ad} \text{ provided } \|\tilde{\mu} - \bar{\mu}\|_2 \text{ sufficiently small}$

Theorem (Kammann/Tröltzsch/V.'13)

 $ar\mu$ optimal control, $ar\mu^\ell$ suboptimal control. Then, second-order sufficient optimality implies

$$\|ar{\mu} - ar{\mu}^{\ell}\|_2 \leq \epsilon_{ape}$$
 with $\epsilon_{ape} = rac{2}{n} \|\zeta(ar{\mu}^{\ell})\|_2$

provided $\|\bar{\mu} - \bar{\mu}^{\ell}\|_2$ is sufficiently small

• Problem: $\eta = \eta(\bar{\mu})$

A-Posteriori Error Analysis (Lass/V.'13)

- Theory: $\|\bar{\mu} \bar{\mu}^{\ell}\|_2 \leq \epsilon_{ape}$ with $\epsilon_{ape} = 2\|\zeta(\bar{\mu}^{\ell})\|_2/\eta$ and $\eta = \lambda_{\min}(\nabla^2 \hat{J}(\bar{\mu}))$
- A-posteriori error:

$$\|\bar{\mu}^{\mathscr{N}} - \bar{\mu}^{\ell}\|_{2} \leq \frac{2}{\lambda_{\min}(\nabla^{2}\hat{J}(\bar{\mu}^{\ell}))} \|\nabla\hat{J}(\bar{\mu}^{\ell})\|_{2} =: \Delta^{\mathscr{N}}(\bar{\mu}^{\ell})$$

• Results (CPU time \approx 30 seconds):

Noise	$\ \bar{\mu}^{\mathscr{N}}-\bar{\mu}^{\ell}\ _{2}$	$\Delta^{\mathscr{N}}(\bar{\mu}^{\ell})$
0%	5.10e-5	8.94e-2
5%	2.53e-4	3.70e-2

• CPU times for the optimization:

	0% noise	5% noise
FE	358.8 seconds	314.6 seconds
POD	9.9 seconds	9.3 seconds

Different approach: utilize Kantorovich theory (Dihlmann/Haasdonk'13)

Related Literature

- Alla, Dihlmann, Haasdonk, Heinkenschloss, Hinze, Manzoni, Quarteroni, Rozza, Sachs, Ulbrich, Ziems...
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