

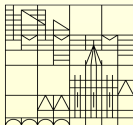
# Surrogate Modeling in PDE Constrained Optimization

Stefan Volkwein

University of Konstanz, Department of Mathematics and Statistics, Numerics & Optimization Group

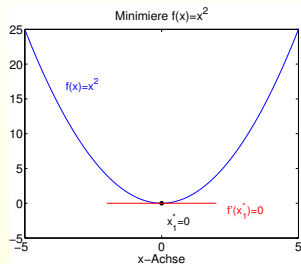
Projection Based Model Reduction:  
RB Methods, POD, and Low Rank Tensor Approximations, November 23-29, 2014

Universität  
Konstanz



## Minimization with Inequality Constraints

- **“Problem”**:  $\min \{f(x) = x^2 : -\infty < x < \infty\}$
- **Optimality condition**:  $f'(x_1^*) \stackrel{!}{=} 0$  with  $f'(x) = 2x$
- **Solution**:  $x_1^* = 0$

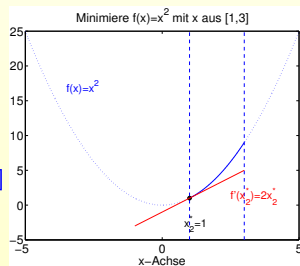


- **Problem**:  $\min \{f(x) = x^2 : 1 \leq x \leq 3\}$
- **Lagrange functional**:

$$L(x, \mu_a, \mu_b) = f(x) + \mu_a(1-x) + \mu_b(x-3)$$

- **Optimality condition**:

$$\begin{pmatrix} f'(x_2^*) - \mu_a^* + \mu_b^* \\ \mu_a^*(1-x_2^*) \\ \mu_b^*(x_2^*-3) \end{pmatrix} \stackrel{!}{=} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \Leftrightarrow \underbrace{f'(x_2^*)(x-x_2^*)}_{2(x-1)} \geq 0 \forall x \in [1,3]$$



- **Solution**:  $x_2^* = 1, \mu_a^* = 2, \mu_b^* = 0$

## Outline of Lecture 2

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

## 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:**

$$\min_{x=(y,u)} J(x) = \frac{1}{2} \|y(t_f) - y_d\|_H^2 + \frac{\kappa}{2} \int_{t_0}^{t_f} \|u(t)\|_U^2 dt$$

subject to the linear evolution problem

$$\frac{d}{dt} \langle y(t), \varphi \rangle_H + a(t; y(t), \varphi) = \langle (f + \mathcal{B}u)(t), \varphi \rangle_{V', V} \quad \forall \varphi \in V \text{ a.e. in } (t_0, t_f]$$

with  $y(t_0) = y_0$  in  $H$  and to bilateral control constraints

$$u \in \mathcal{U}_{ad} = \{ \tilde{u} \in \mathcal{U} \mid u_a(t) \leq \tilde{u}(t) \leq u_b(t) \text{ a.e. in } [t_0, t_f] \}$$

- **State:**  $y(t) \in V \hookrightarrow H$  with Hilbert spaces  $V, H$
- **Control (Hilbert) space:**  $\mathcal{U} = L^2(t_0, t_f; U)$  with  $U = \mathbb{R}^{N_u}$ ,  $U = L^2(\Omega)$  or  $U = L^2(\Gamma)$
- **Input/control:**  $u \in \mathcal{U}_{ad}$  (boundary or distributed control)
- **Bilinear form:**  $a(t; \cdot, \cdot)$  continuous and  $a(t; \varphi, \varphi) \geq \gamma_1 \|\varphi\|_V^2 - \gamma_2 \|\varphi\|_H^2$
- **Control operator:**  $\mathcal{B} : \mathcal{U} \rightarrow L^2(t_0, t_f; V')$  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:**

$$\min_{x=(y,u)} J(x) = \frac{1}{2} \|y(t_f) - y_d\|_H^2 + \frac{\kappa}{2} \int_{t_0}^{t_f} \|u(t)\|_U^2 dt$$

$$\text{s.t. } \frac{d}{dt} \langle y(t), \varphi \rangle_H + a(t; y(t), \varphi) = \langle (f + \mathcal{B}u)(t), \varphi \rangle_{V,V} \quad \forall \varphi \in V \text{ a.e. in } (t_0, t_f]$$

$$y(t_0) = y_0 \text{ in } H \quad \text{and} \quad u_a(t) \leq u(t) \leq u_b(t), \quad t \in [t_0, t_f]$$

- Optimal state  $\bar{y}$ , optimal control  $\bar{u} \in \mathcal{U}_{ad} = \{u \mid u_a \leq u \leq u_b \text{ in } [t_0, t_f]\}$

- **Adjoint/dual equation:**

$$-\frac{d}{dt} \langle \bar{p}(t), \varphi \rangle_H + a(t; \varphi, \bar{p}(t)) = 0 \quad \forall \varphi \in V \text{ a.e. in } [t_0, t_f], \quad \bar{p}(t_f) = \bar{y}(t_f) - y_d$$

- **Variational inequality:**

$$\int_{t_0}^{t_f} \langle \kappa \bar{u}(t) - (\mathcal{B}^* \bar{p})(t), u(t) - \bar{u}(t) \rangle_U dt \geq 0 \quad \forall u \in \mathcal{U}_{ad} \quad (\text{VI})$$

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

$$\langle \hat{J}'(\bar{u}), u(t) - \bar{u}(t) \rangle_{\mathcal{U}} \geq 0 \quad \forall u \in \mathcal{U}_{ad}$$

## POD Galerkin for the State Variable

- **Particular solution:**  $\hat{y} \in W(t_o, t_f)$  solves  $\hat{y}(t_o) = y_o$  in  $H$  and

$$\frac{d}{dt} \langle \hat{y}(t), \varphi \rangle_H + a(t; \hat{y}(t), \varphi) = \langle f(t), \varphi \rangle_{V', V} \quad \forall \varphi \in V \text{ a.e. in } (t_o, t_f]$$

- **POD space:**  $V^\ell = \text{span} \{ \psi_1, \dots, \psi_\ell \} \subset V$
- **POD control-to-state mapping:**  $\mathcal{S}^\ell : \mathcal{U} \rightarrow W(t_o, t_f)$ ,  $w^\ell = \mathcal{S}^\ell u$  solves  $w^\ell(t_o) = 0$  in  $H$  and

$$\frac{d}{dt} \langle w^\ell(t), \psi \rangle_H + a(t; w^\ell(t), \psi) = \langle (\mathcal{B}u)(t), \psi \rangle_{V', V} \quad \forall \psi \in V^\ell \text{ a.e. in } (t_o, t_f]$$

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

- **POD state:**  $y^\ell = \hat{y} + \mathcal{S}^\ell u$
- **Previous theorem with a-priori results for the state:**
  - a)  $y^\ell(0) = y_o$  in  $H$ , i.e., no POD error in the initial condition
  - b)  $\|(\mathcal{S} - \mathcal{S}^\ell)u\|_{W(t_o, t_f)}^2 \leq \sum_{i>\ell} \lambda_i(u)$
  - c)  $\|\mathcal{S} - \mathcal{S}^\ell\|_{\mathcal{L}(\mathcal{U}, W(t_o, t_f))} \rightarrow 0$  for  $\ell \rightarrow \infty$

## POD Galerkin for the Dual Variable (Balancing POD)

- **Adjoint/dual equation:**

$$-\frac{d}{dt} \langle p(t), \varphi \rangle_H + a(t; \varphi, p(t)) = 0 \quad \forall \varphi \in V \text{ a.e. in } [t_0, t_f], \quad p(t_f) = y(t_f) - y_d$$

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

- **Particular solution:**  $\hat{p} \in W(t_0, t_f)$  solves

$$-\frac{d}{dt} \langle \hat{p}(t), \varphi \rangle_H + a(t; \varphi, \hat{p}(t)) = 0 \quad \forall \varphi \in V \text{ a.e. in } [t_0, t_f], \quad \hat{p}(t_f) = \hat{y}(t_f) - y_d$$

- **Dual solution operator:**  $\mathcal{A} : \mathcal{U} \rightarrow W(t_0, t_f)$ ,  $v = \mathcal{A}u$  solves

$$-\frac{d}{dt} \langle v(t), \varphi \rangle_H + a(t; \varphi, v(t)) = 0 \quad \forall \varphi \in V \text{ a.e. in } [t_0, t_f], \quad v(t_f) = (\mathcal{S}u)(t_f)$$

- **POD dual solution operator:**  $\mathcal{A}^\ell : \mathcal{U} \rightarrow W(t_0, t_f)$ ,  $v^\ell = \mathcal{A}^\ell u$  solves

$$-\frac{d}{dt} \langle v^\ell(t), \psi \rangle_H + a(t; \psi, v^\ell(t)) = 0 \quad \forall \psi \in V^\ell \text{ a.e. in } [t_0, t_f], \quad v^\ell(t_f) = (\mathcal{S}^\ell u)(t_f) \in V^\ell$$

⇒ 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

$$\min \sum_{k=1}^{\mathcal{P}} \int_{t_0}^{t_f} \left\| y^k(t) - \sum_{i=1}^{\ell} \langle y^k(t), \psi_i \rangle_X \psi_i \right\|_X^2 dt \text{ s.t. } \{ \psi_i \}_{i=1}^{\ell} \subset X, \langle \psi_i, \psi_j \rangle_X = \delta_{ij}, 1 \leq i, j \leq \ell \quad (\mathbf{P}^{\ell})$$

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

$X = V$ ,  $\mathcal{V} = \text{span} \{ y^k(t) \mid t \in [t_0, t_f], 1 \leq k \leq \mathcal{P} \}$

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

$$\|p - p^\ell\|_{W(t_0, t_f)}^2 \leq C_1 \left( \sum_{i>\ell} \lambda_i + \|y_1^1 - \mathcal{P}^\ell y_1^1\|_{L^2(t_0, t_f; V)}^2 + \|y_1^2 - \mathcal{P}^\ell y_1^2\|_{L^2(t_0, t_f; V)}^2 \right)$$

b) Snapshots  $y^1 = \mathcal{S}u$ ,  $y^2 = \mathcal{A}u$ ,  $y^3 = (\mathcal{S}u)_t$ ,  $y^4 = (\mathcal{A}u)_t$ , all in  $L^2(t_0, t_f; V)$ :

$$\|p - p^\ell\|_{W(t_0, t_f)}^2 \leq C_2 \sum_{i>\ell} \lambda_i$$

c) If  $\mathcal{S}\tilde{u}, \mathcal{A}\tilde{u} \in H^1(t_0, t_f; V)$  for all  $\tilde{u} \in \mathcal{U}$ , then  $\lim_{\ell \rightarrow \infty} \|\mathcal{A} - \mathcal{A}^\ell\|_{\mathcal{L}(\mathcal{U}, W(t_0, t_f))} = 0$

In particular,  $\lim_{\ell \rightarrow \infty} \|p(\tilde{u}) - p^\ell(\tilde{u})\|_{W(t_0, t_f)} = 0$  for any  $\tilde{u} \in \mathcal{U}$

## POD Approximation of the Variational Inequality

- **Variational inequality:**

$$\int_{t_0}^{t_f} \langle \kappa \bar{u}(t) - (\mathcal{B}^* \bar{p})(t), u(t) - \bar{u}(t) \rangle_U dt \geq 0 \quad \forall u \in \mathcal{U}_{ad} \quad (\text{VI})$$

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

- **POD variational inequality:**

$$\int_{t_0}^{t_f} \langle \kappa \bar{u}^\ell(t) - (\mathcal{B}^* \bar{p}^\ell)(t), u(t) - \bar{u}^\ell(t) \rangle_U dt \geq 0 \quad \forall u \in \mathcal{U}_{ad} \quad (\text{VI}^\ell)$$

- **A-priori analysis:** choose  $u = \bar{u}^\ell$  in (VI),  $u = \bar{u}$  in (VI<sup>ℓ</sup>) and add

$$\begin{aligned} 0 &\leq \int_{t_0}^{t_f} \langle \kappa(\bar{u} - \bar{u}^\ell)(t) - (\mathcal{B}^*(\bar{p} - \bar{p}^\ell))(t), \bar{u}^\ell(t) - \bar{u}(t) \rangle_U dt \\ &= -\kappa \|\bar{u} - \bar{u}^\ell\|_{\mathcal{U}}^2 - \underbrace{\int_{t_0}^{t_f} \langle (\mathcal{B}^*(\bar{p} - \bar{p}^\ell))(t), \bar{u}^\ell(t) - \bar{u}(t) \rangle_U dt}_{\leq C \|(\mathcal{A} - \mathcal{A}^\ell)\bar{u}\|_{L^2(t_0, t_f; V)} \|\bar{u}^\ell - \bar{u}\|_{\mathcal{U}}} \end{aligned}$$

## Convergence Result for the POD Suboptimal Control

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

$$\min \sum_{k=1}^{\wp} \int_{t_0}^{t_f} \left\| y^k(t) - \sum_{i=1}^{\ell} \langle y^k(t), \psi_i \rangle_X \psi_i \right\|_X^2 dt \text{ s.t. } \{\psi_i\}_{i=1}^{\ell} \subset X, \langle \psi_i, \psi_j \rangle_X = \delta_{ij}, 1 \leq i, j \leq \ell \quad (\mathbf{P}_{\infty}^{\ell})$$

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

$X = V$ ,  $\mathcal{Y} = \text{span}\{y^k(t) \mid t \in [t_0, t_f], 1 \leq k \leq \wp\}$

a) Snapshots  $y^1 = \mathcal{S}\bar{u}$ ,  $y^2 = \mathcal{A}\bar{u}$ ,  $y^3 = (\mathcal{S}\bar{u})_t$ ,  $y^4 = (\mathcal{A}\bar{u})_t$ , all in  $L^2(t_0, t_f; V)$ :

$$\|\bar{u} - \bar{u}^{\ell}\|_{W(t_0, t_f)}^2 \leq C_2 \sum_{i>\ell} \lambda_i$$

b) If  $\mathcal{S}\tilde{u}, \mathcal{A}\tilde{u} \in H^1(t_0, t_f; V)$  for all  $\tilde{u} \in \mathcal{U}$ , then  $\lim_{\ell \rightarrow \infty} \|\bar{u} - \bar{u}^{\ell}\|_{\mathcal{U}} = 0$

# Perturbation Analysis (Malanowski/Büskens/Maurer'97)

- **Variational inequality:**

$$\int_{t_0}^{t_f} \langle \kappa \bar{u}(t) - (\mathcal{B}^* \bar{p})(t), u(t) - \bar{u}(t) \rangle_U dt \geq 0 \quad \forall u \in \mathcal{U}_{ad} \quad (\text{VI})$$

- **Misfit in the variational inequality:** suboptimal  $\bar{u}^\ell \in \mathcal{U}_{ad}$

$$\int_{t_0}^{t_f} \langle \kappa \bar{u}^\ell(t) - (\mathcal{B}^* \tilde{p}^\ell)(t), u(t) - \bar{u}^\ell(t) \rangle_U dt \not\geq 0 \quad \forall u \in \mathcal{U}_{ad}$$

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

- **Perturbation analysis:** there exists a perturbation  $\zeta^\ell \in \mathcal{U}$  satisfying

$$\int_{t_0}^{t_f} \langle \kappa \bar{u}^\ell(t) - (\mathcal{B}^* \tilde{p}^\ell)(t) + \zeta^\ell(t), u(t) - \bar{u}^\ell(t) \rangle_U dt \geq 0 \quad \forall u \in \mathcal{U}_{ad} \quad (\tilde{\text{VI}}^\ell)$$

- **A-posteriori analysis:** choose  $u = \bar{u}^\ell$  in (VI),  $u = \bar{u}$  in ( $\tilde{\text{VI}}^\ell$ ) and add

$$\kappa \|\bar{u} - \bar{u}^\ell\|_{\mathcal{U}}^2 \leq \int_{t_0}^{t_f} \langle (\mathcal{B}^* \mathcal{A}(\bar{u}^\ell - \bar{u})) + \zeta^\ell(t), \bar{u}^\ell(t) - \bar{u}(t) \rangle_U dt$$

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

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

## Convergence Result for the Perturbation

- **Estimate for the control:**  $\|\bar{u} - \bar{u}^\ell\|_{\mathcal{U}} \leq \frac{1}{\kappa} \|\zeta^\ell\|_{\mathcal{U}}$  and  $\tilde{p}^\ell = \hat{p} + \mathcal{A}\bar{u}^\ell$
  - **Computation of  $\zeta^\ell$ :**  $\zeta^\ell(t) = \begin{cases} -(\kappa\bar{u}^\ell(t) - (\mathcal{B}^*\tilde{p}^\ell)(t)) & \text{if } u_a(t) < \bar{u}^\ell(t) < u_b(t) \\ -\min(0, \kappa\bar{u}^\ell(t) - (\mathcal{B}^*\tilde{p}^\ell)(t)) & \text{if } \bar{u}^\ell(t) = u_a(t) \\ -\max(0, \kappa\bar{u}^\ell(t) - (\mathcal{B}^*\tilde{p}^\ell)(t)) & \text{if } \bar{u}^\ell(t) = u_b(t) \end{cases}$
- i.e.,  $\zeta^\ell = \zeta^\ell(\bar{u}^\ell) \Rightarrow$  a-posteriori error analysis for suboptimal  $\bar{u}^\ell$

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

$X = V$ ,  $\mathcal{Y} = \text{span}\{y^k(t) \mid t \in [t_0, t_f], 1 \leq k \leq \rho\}$

a) Snapshots  $y^1 = \mathcal{J}\bar{u}$ ,  $y^2 = \mathcal{A}\bar{u}$ ,  $y^3 = (\mathcal{J}\bar{u})_t$ ,  $y^4 = (\mathcal{A}\bar{u})_t$ , all in  $L^2(t_0, t_f; V)$ :

$$\|\zeta^\ell\|_{\mathcal{U}}^2 \leq C \sum_{i>\ell} \lambda_i$$

b) If  $\mathcal{J}\tilde{u}, \mathcal{A}\tilde{u} \in H^1(t_0, t_f; V)$  for all  $\tilde{u} \in \mathcal{U}$ , then  $\lim_{\ell \rightarrow \infty} \|\zeta^\ell\|_{\mathcal{U}} = 0$

## Algorithm with POD A-Posteriori Analysis

- **Estimate for the control:**  $\|\bar{u} - \bar{u}^\ell\|_{\mathcal{U}} \leq \frac{1}{\kappa} \|\zeta^\ell\|_{\mathcal{U}}$  and  $\tilde{p}^\ell = \hat{p} + \mathcal{A}\bar{u}^\ell$
- **Computation of  $\zeta^\ell$ :** 
$$\zeta^\ell(t) = \begin{cases} -(\kappa\bar{u}^\ell(t) - (\mathcal{B}^*\tilde{p}^\ell)(t)) & \text{if } u_a(t) < \bar{u}^\ell(t) < u_b(t) \\ -\min(0, \kappa\bar{u}^\ell(t) - (\mathcal{B}^*\tilde{p}^\ell)(t)) & \text{if } \bar{u}^\ell(t) = u_a(t) \\ -\max(0, \kappa\bar{u}^\ell(t) - (\mathcal{B}^*\tilde{p}^\ell)(t)) & \text{if } \bar{u}^\ell(t) = u_b(t) \end{cases}$$

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

- (1) Choose  $\ell_{\max}$  and POD basis  $\{\psi_i\}_{i=1}^{\ell_{\max}}$  for the Galerkin approximation of the LQ problem;
- (2) Determine the reduced-order model for the LQ problem;
- (3) Calculate suboptimal control  $\bar{u}^\ell \in \mathcal{U}_{ad}$ , e.g., by a semismooth Newton method;
- (4) Compute perturbation  $\bar{\zeta}^\ell = \zeta^\ell(\bar{u}^\ell)$ ;
- (5) **IF**  $\|\bar{\zeta}^\ell\|_{\mathcal{U}} / \kappa > \text{TOL}$  **AND**  $\ell < \ell_{\max}$  **THEN**  
     Enlarge  $\ell$  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)

## Numerical Example: Problem Formulation and Optimal Control (Stuedinger'12, Stuedinger/V.'13)

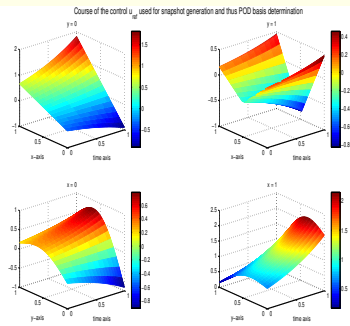
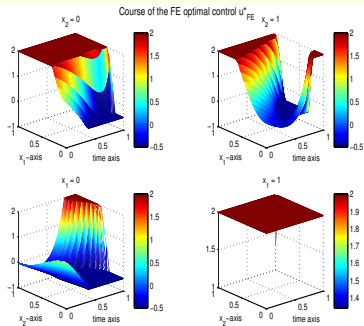
## ● Consider:

$$\min J(y, u) = \frac{1}{2} \int_{\Omega} |y(t_f) - y^d|^2 dx + \frac{1}{200} \int_0^{t_f} \int_{\Gamma} |u|^2 dx dt$$

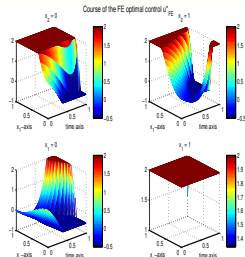
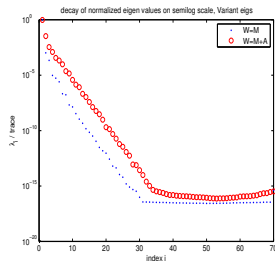
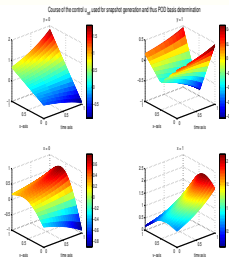
$$\text{s.t. } y_t + 0.1 \Delta y = 0 \text{ in } \mathcal{Q}, \quad \frac{\partial y}{\partial n} + \frac{y}{100} = u \text{ on } \Sigma, \quad y(0) = y_0 \text{ in } \Omega = (0, 1)^2$$

$$-0.5 \leq u \leq 2 \text{ on } \Sigma = (0, t_f) \times \Gamma$$

## ● Method &amp; Discretization: semismooth Newton &amp; implicit Euler, finite elements



## Numerical Example: POD Error Analysis (Stuedinger'12, Stuedinger/V.'13)



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

$\ell$	$\varepsilon_{ape}$	$\ \bar{u}^h - \bar{u}^\ell\ $	$\frac{\varepsilon_{ape}}{\ \bar{u} - \bar{u}^\ell\ }$	$\varepsilon_{ape}$	$\ \bar{u}^h - \bar{u}^\ell\ $	$\frac{\varepsilon_{ape}}{\ \bar{u} - \bar{u}^\ell\ }$
5	$1.3 \cdot 10^{-0}$	$9.1 \cdot 10^{-1}$	1.32	$6.5 \cdot 10^{-1}$	$5.6 \cdot 10^{-1}$	1.16
20	$5.9 \cdot 10^{-1}$	$3.2 \cdot 10^{-1}$	1.84	$7.5 \cdot 10^{-3}$	$7.3 \cdot 10^{-3}$	1.03
60	$1.4 \cdot 10^{-2}$	$1.2 \cdot 10^{-2}$	1.17	$8.3 \cdot 10^{-5}$	$8.3 \cdot 10^{-5}$	1.00
70	$1.2 \cdot 10^{-2}$	$1.1 \cdot 10^{-2}$	1.10	$3.0 \cdot 10^{-5}$	$3.0 \cdot 10^{-5}$	1.00
90	$1.1 \cdot 10^{-2}$	$9.7 \cdot 10^{-3}$	1.13	$3.7 \cdot 10^{-6}$	$3.7 \cdot 10^{-6}$	1.00

**Problem:** optimal FE control  $\bar{u}^h$  is unknown  $\Rightarrow$  **POD basis update**



## Optimality-System POD (OS-POD) (Kunisch/V.'08)

● **Original problem:**  $\min \hat{J}(u) = J(y(u), u)$  s.t.  $u \in \mathcal{U}_{ad}$

● **POD-Galerkin approximation:**

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

● **OSPOD-Problem:**

$$\min \hat{\mathcal{J}}^\ell(u) = J(y^\ell(u), u) \quad \text{s.t.} \quad \begin{cases} y = y(u), & \psi_i = \psi_i(u) \text{ for } 1 \leq i \leq \ell \\ \mathcal{R} \psi_i = \int_{t_0}^{t_f} \langle y(t), \psi_i \rangle_X y(t) dt = \lambda_i \psi_i, & 1 \leq i \leq \ell \end{cases} \quad (\hat{\mathbf{P}}_{\text{ospod}}^\ell)$$

→ 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  $(\hat{\mathbf{P}}_{\text{ospod}}^\ell)$
- compute an approximate solution to  $(\hat{\mathbf{P}}^\ell)$  by **Algorithm 1**

● **Efficient combination with a-posteriori error estimator** (Grimm'13, Grimm/Gubisch/V.'14, V.'11)

Algorithm for A-Posteriori Error with OS-POD (Grimm'13, Grimm/Gubisch/V.'14, V.'11)

● **OS-POD-Problem:**

$$\min \hat{J}^\ell(u) = J(y^\ell(u), u) \quad \text{s.t.} \quad \begin{cases} y = y(u), & \psi_i = \psi_i(u) \text{ for } 1 \leq i \leq \ell \\ \mathcal{R}\psi_i = \int_{t_0}^{t_f} \langle y(t), \psi_i \rangle_X y(t) dt = \lambda_i \psi_i, & 1 \leq i \leq \ell \end{cases} \quad (\hat{\mathbf{P}}_{\text{ospod}}^\ell)$$

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

- (1) Choose initial control  $u^0$  and number of basis elements  $\ell < \ell_{\max}$ ;
- (2) Perform  $k$  OS-POD gradient steps to get  $u^k$  and  $\{\psi_i(u^k)\}_{i=1}^\ell$ ;

---

- (3) Determine the reduced-order model for the LQ problem;
- (4) Calculate suboptimal control  $\bar{u}^\ell \in \mathcal{U}_{\text{ad}}$ , e.g., by a semismooth Newton method;
- (5) Compute perturbation  $\bar{\zeta}^\ell = \zeta^\ell(\bar{u}^\ell)$ ;
- (6) **IF**  $\|\bar{\zeta}^\ell\|_{\mathcal{U}} / \kappa > \text{TOL}$  **AND**  $\ell < \ell_{\max}$  **THEN**  
     Enlarge  $\ell$  and go back to (3);  
   **ELSE**  
     Stop;  
   **ENDIF**

- **Algorithm:** apply a few gradient steps for  $(\hat{\mathbf{P}}_{\text{ospod}}^\ell)$  and continue with fixed POD basis
- **Stopping criterium:**  $\|\bar{\zeta}^\ell\|_{\mathcal{U}} / \kappa = \mathcal{O}(\Delta t^\alpha + \Delta x^\beta)$

## OS-POD for linear-quadratic, control constrained control (Grimm'13, Grimm/Gubisch/V.'14, V.'11)

● **Consider:**

$$\min J(y, u) = \frac{1}{2} \int_{\Omega} |y(t_f) - y^d|^2 dx + \frac{1}{20} \int_0^{t_f} \int_{\Gamma} |u|^2 dx dt$$

$$\text{s.t. } y_t + 0.1 \Delta y = 0 \text{ in } \mathcal{Q}, \quad \frac{\partial y}{\partial n} + \frac{y}{100} = u \text{ on } \Sigma, \quad y(0) = y_0 \text{ in } \Omega = (0, 1)^2$$

$$-0 \leq u \leq 1 \text{ on } \Sigma = (0, t_f) \times \Gamma$$

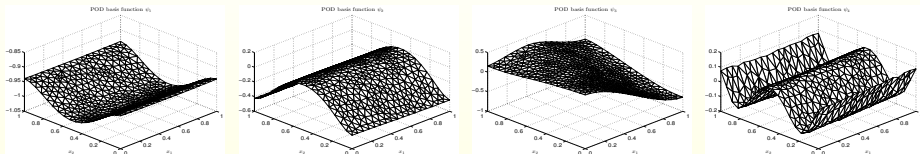
● **Method & Discretization:** semismooth Newton & implizit Euler, finite elements

	$k=0$	$k=1$	$k=2$	with $u_{FE}$
required $\ell$	35	40	13	13
CPU	110.77 sec	147.14 sec	18.39 sec	11.48 sec
$\ \zeta^{\ell}\ /\sigma_u$	$3.43 \cdot 10^{-3}$	$1.14 \cdot 10^{-2}$	$2.82 \cdot 10^{-3}$	$1.94 \cdot 10^{-3}$
$e_{abs}^u$	$3.15 \cdot 10^{-3}$	$9.53 \cdot 10^{-3}$	$2.62 \cdot 10^{-3}$	$1.93 \cdot 10^{-3}$
$e_{rel}^u$	$2.45 \cdot 10^{-3}$	$7.73 \cdot 10^{-3}$	$2.15 \cdot 10^{-3}$	$1.59 \cdot 10^{-3}$
$e_{abs}^y$	$1.59 \cdot 10^{-2}$	$3.10 \cdot 10^{-3}$	$7.55 \cdot 10^{-4}$	$1.92 \cdot 10^{-4}$
$e_{rel}^y$	$5.75 \cdot 10^{-3}$	$1.12 \cdot 10^{-3}$	$2.73 \cdot 10^{-4}$	$6.97 \cdot 10^{-5}$

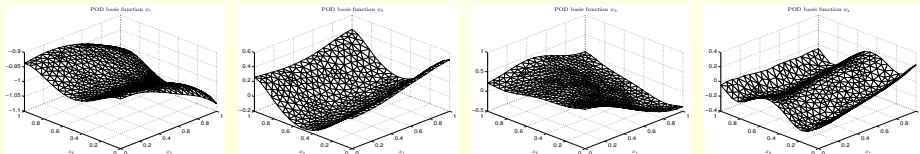
	$k=0$	$k=1$	$k=2$	with $u_{FE}$
$\ \zeta^{15}\ /\sigma_u$	$2.50 \cdot 10^{-2}$	$1.45 \cdot 10^{-2}$	$2.27 \cdot 10^{-3}$	$1.59 \cdot 10^{-3}$
$e_{abs}^u$	$2.06 \cdot 10^{-2}$	$1.19 \cdot 10^{-2}$	$2.07 \cdot 10^{-3}$	$1.59 \cdot 10^{-3}$
$e_{rel}^u$	$1.65 \cdot 10^{-2}$	$9.65 \cdot 10^{-3}$	$1.67 \cdot 10^{-3}$	$1.30 \cdot 10^{-3}$
$e_{abs}^y$	$2.59 \cdot 10^{-2}$	$6.34 \cdot 10^{-3}$	$6.42 \cdot 10^{-4}$	$6.91 \cdot 10^{-5}$
$e_{rel}^y$	$9.34 \cdot 10^{-3}$	$2.29 \cdot 10^{-3}$	$2.32 \cdot 10^{-4}$	$2.51 \cdot 10^{-5}$
different $u_a$	96	67	15	16(2233)
different $u_b$	63	38	6	4(3891)

## Convergence of the POD basis (Grimm'13, Grimm/Gubisch/V.'14, Kunisch/Müller'14)

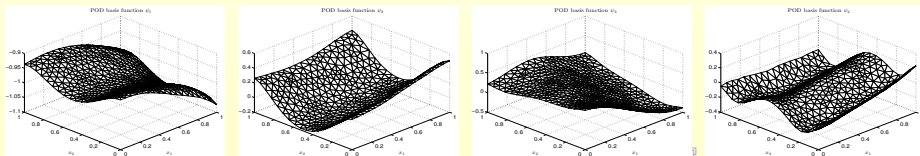
- First four POD basis functions generated from  $u = 0$ :



- First four POD basis functions generated from the optimal control  $\bar{u}^h$ :



- First four POD basis functions generated after  $k = 2$  OS-POD gradient steps:



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

- **(Abstract) linear-quadratic problem:**

$$\min_{x=(y,u)} J(x) = \frac{1}{2} \|y(t_f) - y_d\|_H^2 + \frac{\kappa}{2} \int_{t_0}^{t_f} \|u(t)\|_U^2 dt$$

$$\text{s.t. } \frac{d}{dt} \langle y(t), \varphi \rangle_H + a(t; y(t), \varphi) = \langle (f + \mathcal{B}u)(t), \varphi \rangle_{V',V} \quad \forall \varphi \in V \text{ a.e. in } (t_0, t_f]$$

$$y(t_0) = y_0 \text{ in } H \quad \text{and} \quad u_a(t) \leq \varepsilon u(t) + (\mathcal{I}y)(t) \leq u_b(t), \quad t \in [t_0, t_f]$$

- **Control space:**  $\mathcal{U} = L^2(t_0, t_f; U)$  with  $U = \mathbb{R}^{N_u}$
- **Input/control operator:**  $(\mathcal{B}u)(t, x) = \sum_{i=1}^{N_u} u_i(t) \chi_{\Omega_i}(x)$
- **State operator:**  $(\mathcal{I}y)(t) = (\int_{\Omega_i} y(t, x) dx / |\Omega_i|)_{1 \leq i \leq N_u}$
- **State constraints:**  $\varepsilon \rightarrow 0$
- **Particular solution:**  $\hat{y} \in W(t_0, t_f)$  solves  $\hat{y}(t_0) = y_0$  in  $H$  and

$$\frac{d}{dt} \langle \hat{y}(t), \varphi \rangle_H + a(t; \hat{y}(t), \varphi) = \langle f(t), \varphi \rangle_{V',V} \quad \forall \varphi \in V \text{ a.e. in } (t_0, t_f]$$

- **Control-to-state mapping:**  $\mathcal{S} : \mathcal{U} \rightarrow W(t_0, t_f)$ ,  $w = \mathcal{S}u$  solves  $w(t_0) = 0$  in  $H$  and

$$\frac{d}{dt} \langle w(t), \varphi \rangle_H + a(t; w(t), \varphi) = \langle (\mathcal{B}u)(t), \varphi \rangle_{V',V} \quad \forall \varphi \in V \text{ a.e. in } (t_0, t_f]$$

- **Inequality constraints:**  $v_a(t) := (u_a - \mathcal{I}\hat{y})(t) \leq (\varepsilon u + \mathcal{I}\mathcal{S}u)(t) \leq (u_b - \mathcal{I}\hat{y}) := v_b(t)$

## Formulation as a Control Constrained Problem

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

$$\min_v J(\hat{y} + \mathcal{I}\mathcal{F}^{-1}v, \mathcal{F}^{-1}v) = \dots + \frac{\kappa}{2} \|\mathcal{F}^{-1}v\|^2$$

$$\text{s.t. } v \in V_{\text{ad}} = \{\tilde{v} \mid v_a(t) \leq \tilde{v}(t) \leq v_b(t) \text{ for } t \in [t_0, t_f]\}$$

⇒ form of the previous linear-quadratic optimal control problem

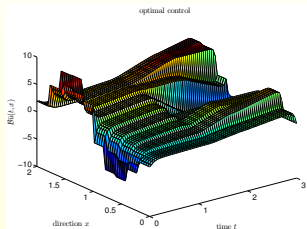
- **Variational inequality:**

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

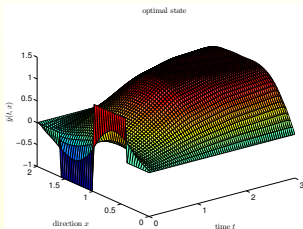
- **POD Galerkin scheme:** additional analysis for  $\mathcal{F}^\ell \approx \mathcal{F}$
- **A-posteriori error estimate** (Gubisch/V'14):  $\|\bar{u} - \bar{u}^\ell\| \leq \frac{1}{\kappa} \|\mathcal{F}\|_{\mathcal{L}(\mathcal{U})} \|\zeta^\ell\|$  with computable  $\zeta^\ell$
- **Convergence result** (Gubisch/V'14):  $\|\zeta^\ell\| \rightarrow 0$  for  $\ell \rightarrow \infty$

Numerical Example: POD basis update (Afanasiev/Hinze'01, Gubisch/V'14)

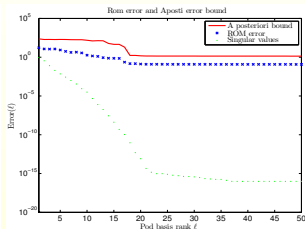
- **Optimal control problem:** heat equation, control space  $\mathcal{U} = L^2(0,3;\mathbb{R}^{10})$ ,  $\varepsilon = 1e-5$



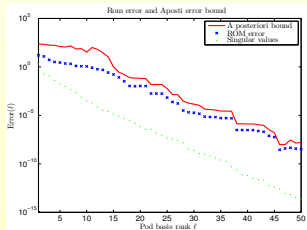
Optimal control



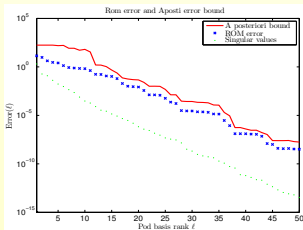
Optimal state



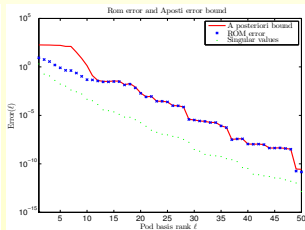
POD basis with  $u = 1$



POD basis update



POD basis with  $u = \tilde{u}$



POD basis update ( $\varepsilon = 1e-3$ )

## State and Control Constrained Linear-Quadratic Optimal Control Problem

- **Quadratic programming (QP) problem:**

$$\min_{x=(y,u)} J(x) = \frac{1}{2} \|y(t_f) - y_d\|_{L^2(\Omega)}^2 + \frac{\kappa}{2} \int_{t_0}^{t_f} \|u(t)\|_U^2 dt$$

subject to the linear evolution problem

$$\langle y_t(t), \varphi \rangle + a(t; y(t))(t, \varphi) = \langle (f + \mathcal{B}u)(t), \varphi \rangle \quad \forall \varphi \in V \text{ in } (t_0, t_f]$$

with  $y(t_0) = y_0$  and to bilateral control constraints

$$u \in \mathcal{U}_{ad} = \{v \in \mathcal{U} \mid u_a(t) \leq v(t) \leq u_b(t) \text{ in } [t_0, t_f]\}$$

$$y \in \{z \in L^2(Q) \mid y_a(t, \mathbf{x}) \leq z(t, \mathbf{x}) \leq y_b(t, \mathbf{x}) \text{ in } Q\}$$

- **Lavrentiev regularization:**  $\varepsilon > 0$  and  $x = (y, u)$

$$J(x, w) = \frac{1}{2} \|y(t_f) - y_d\|_{L^2(\Omega)}^2 + \frac{\kappa}{2} \int_{t_0}^{t_f} \|u(t)\|_U^2 dt + \frac{\sigma}{2} \int_{t_0}^{t_f} \|w(t)\|_{L^2(\Omega)}^2 dt$$

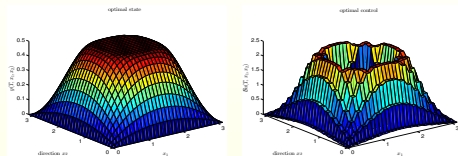
$$(y, w) \in \{(z, v) \in L^2(Q) \times L^2(Q) \mid y_a(t, \mathbf{x}) \leq \varepsilon v(t, \mathbf{x}) + z(t, \mathbf{x}) \leq y_b(t, \mathbf{x}) \text{ in } Q\}$$

- **Regular Lagrange multipliers** (Tröltzsch'05): formulation as a control constrained problem
- **A-posteriori error** (Gubisch'14): extension of the analysis from the control-constrained case  $\Rightarrow$  OS-POD also applicable

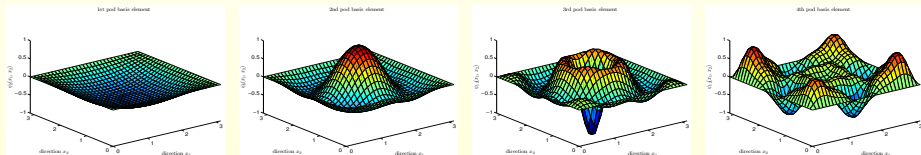


## Numerical Example: State and Control Constraints (Gubisch '14, Grimm/Gubisch/V'14)

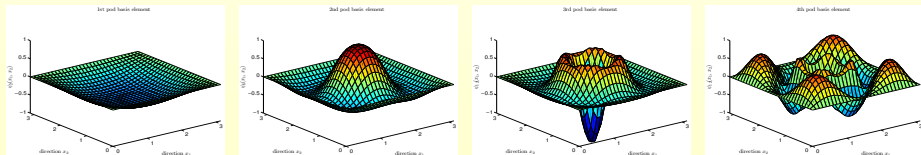
- Two-dimensional heat equation
- Distributed control
- Semismooth Newton method



- First four POD basis functions generated from the optimal control  $\bar{u}$ :



- First four POD basis functions generated after three OS-POD gradient steps:



## Related Literature

- Afanasiev, Dede, Fahl, Grepl, Hinze, Kärcher, Manzoni, Negri, Quarteroni, Rozza, Sachs, Schu...
- E. Grimm: *Optimality system POD and a-posteriori error analysis for linear-quadratic optimal control problems*, 2013
- E. Grimm, M. Gubisch, S. V.: *Numerical analysis of optimality-system POD for constrained optimal control*, 2014
- M. Gubisch, S. V.: *POD a-posteriori error analysis for optimal control problems with mixed control-state constraints*, 2014
- M. Hinze, S. V.: *Error estimates for abstract linear-quadratic optimal control problems using POD*, 2008
- K. Kunisch, S. V.: *POD for optimality systems*, 2010
- A. Studinger, S.V.: *Numerical analysis of POD a-posteriori error estimation for optimal control*, 2013
- T. Tonn, K. Urban, S.V.: *Comparison of the reduced-basis and POD a-posteriori error estimators for an elliptic linear-quadratic optimal control problem*, 2011
- F. Tröltzsch, S. V.: *POD a-posteriori error estimates for linear-quadratic optimal control problems*, 2009
- S.V.: *Optimality system POD and a-posteriori error analysis for linear-quadratic problems*, 2011

## Extensions to Nonlinear PDE Constrained Optimization

### Topics:

- Inexact sequential quadratic programming (SQP)
- A-posteriori error analysis
- Trust-region POD

## Sequential Quadratic Programming (SQP)

- **Infinite dimensional optimization:**

$$\min J(x) \quad \text{s.t.} \quad e(x) = 0, \quad g(x) \leq 0 \quad (\mathbf{P})$$

- **Lagrange functional for (P):**  $\mathcal{L}(x, p, q) = J(x) + \langle e(x), p \rangle + \langle g(x), q \rangle$

- **(Local) SQP method:** at  $z_k = (x_k, p_k, q_k)$  solve

$$\begin{cases} \min_{x_\delta} \mathcal{L}_x(z_k)x_\delta + \frac{1}{2} \mathcal{L}_{xx}(z_k)(x_\delta, x_\delta) \\ \text{s.t. } e(x_k) + e'(x_k)x_\delta = 0, \quad g(x_k) + g'(x_k)x_\delta \leq 0 \end{cases} \quad (\mathbf{QP}^k)$$

- **KKT system for equality constraints:** solution  $\bar{x}_\delta$  to  $(\mathbf{QP}^k)$  is characterized by

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

## Inexact SQP by Using POD or RB

- **KKT system:** solution  $\bar{x}_\delta$  to  $(\mathbf{QP}^k)$  is characterized by

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

- **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:**  $\bar{z}_\delta^\ell$  reduced-order solution

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

with prolongation  $\mathcal{P}$

- **Rate of convergence:** superlinear ( $1 < q < 2$ ), quadratic ( $q = 2$ )

- **Control of reduced-order approach:**

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

## Convergence Result

● **Variables in optimal control:**  $x = (y, u)$ ,  $y = y(u)$

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

$$\left( \begin{array}{cc|c} \mathcal{L}_{yy}(z_k) & \mathcal{L}_{yu}(z_k) & e_y(x_k)^* \\ \mathcal{L}_{uy}(z_k) & \mathcal{L}_{uu}(z_k) & e_u(x_k)^* \\ \hline e_y(x_k) & e_u(x_k) & 0 \end{array} \right) \begin{pmatrix} y_\delta \\ u_\delta \\ p_\delta \end{pmatrix} = \begin{pmatrix} -\mathcal{L}_y(z_k) \\ -\mathcal{L}_u(z_k) \\ -e(x_k) \end{pmatrix}$$

● **Suboptimal solution to KKT system:**  $\bar{z}_\delta^\ell = (\bar{y}_\delta^\ell, \bar{u}_\delta^\ell, \bar{p}_\delta^\ell)$

● **Prolongation**  $\mathcal{P}: \bar{z}_\delta^\ell \mapsto \mathcal{P}\bar{z}_\delta^\ell = (\tilde{y}_\delta, \tilde{u}_\delta, \tilde{p}_\delta)$  with

$$e_y(x_k)\tilde{y}_\delta = -e(x_k) - e_u(x_k)\tilde{u}_\delta$$

$$e_y(x_k)^*\tilde{p}_\delta = -\mathcal{L}_y(z_k) - \mathcal{L}_{yy}(z_k)\tilde{y} - \mathcal{L}_{yu}(z_k)\tilde{u}_\delta$$

→ a-posteriori error computable

**Theorem** (Kahlbacher/V. '12)

*Second-order sufficient optimality implies*

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

## Multilevel Approach with Reduced-Order Models

- **Convergence criterium:**  $\|A_k \mathcal{P} \bar{z}_\delta^\ell - b_k\| \simeq \|\bar{u}_\delta - \bar{u}_\delta^\ell\| < \text{TOL}$

- **A-posteriori error** (Tröltzsch/V.'09):

$$\|\bar{u}_\delta - \bar{u}_\delta^\ell\| \simeq \underbrace{\|\mathcal{L}_{uy}(z_k) \tilde{y}_\delta + \mathcal{L}_{uu}(z_k) \bar{u}_\delta^\ell + e_u(x_k)^* \tilde{p}_\delta + \mathcal{L}_u(z_k)\|}_{:= -\bar{\zeta}^\ell}$$

with  $\|\bar{\zeta}^\ell\| \rightarrow 0$  for  $\ell \rightarrow \infty$

- **Convergence of  $\|\bar{\zeta}^\ell\|$ :** no rate, basis dependent (Hinze/V.'08)
- **POD basis:** combination with **Optimality-System POD** (Metzdorf'15)
- **Combination with FE adaptivity:** (Clever/Lang/Ulbrich/Ziems)

- **Reduced approach:**  $\min \hat{J}(u) = J(y(u), u)$

- **QP problem:** for given iterate  $u^k$  solve

$$\min q^k(u_\delta) = \hat{J}(u^k) + \hat{J}'(u^k)u_\delta + \frac{1}{2} \hat{J}''(u^k)(u_\delta, u_\delta)$$

- **Globalization by inexact trust-region:** inexact cost  $q_\ell^k \approx q^k$  and inexact solution of

$$\min q_\ell^k(u_\delta) = \hat{J}_\ell(u^k) + \hat{J}'_\ell(u^k)u_\delta + \frac{1}{2} \hat{J}''_\ell(u^k)(u_\delta, u_\delta) \quad \text{s.t.} \quad \|u_\delta\| \leq \Delta_k$$

- **Convergence criterium:** Carter condition  $\|\hat{J}'(u^k) - \hat{J}'_\ell(u^k)\| \leq \zeta \|\hat{J}'_\ell(u^k)\|$  with  $\zeta \in (0, 1)$
- **Trust-region POD:** (Arian/Fahl/Sachs'00, Schu/Sachs'12) and (Rogg'14)

## Related Literature

- Alla, Dihlmann, Haasdonk, Heinkenschloss, Hinze, Manzoni, Quarteroni, Rozza, Sachs, Ulbrich, Ziems...
- P. Benner, E.W. Sachs, S. V.: *Model order reduction for PDE constrained optimization*, 2014
- C.-C. Gräßle: *POD based inexact SQP methods for optimal control problems governed by a semilinear heat equation*, 2014
- M. Kahlbacher, S.V. *POD a-posteriori error based inexact SQP method for bilinear elliptic optimal control problems*, 2012
- E. Kammann, F. Tröltzsch, S.V.: *A method of a-posteriori error estimation with application to POD*, 2013
- O. Lass, S. V.: *Parameter identification for nonlinear elliptic-parabolic systems with application in lithium-ion battery modeling*, 2014
- S. Rogg: *Trust region POD for optimal boundary control of a semilinear heat equation*, 2014
- E.W. Sachs, S.V.: *POD-Galerkin approximations in PDE-constrained optimization*, 2010
- S.V.: *Proper Orthogonal Decomposition zur Optimalsteuerung linearer partieller Differentialgleichungen*, 2013