

# Efficient Optimal Control of Young Concrete Mechanical Properties

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Workshop on PDE Constrained  
Optimization of Certain and  
Uncertain Processes,

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- 1 Young Concrete. Problem Introduction and Modelling
- 2 State Constrained Parabolic Optimal Control Problem
- 3 A-posteriori Error Estimation
- 4 Numerical Results

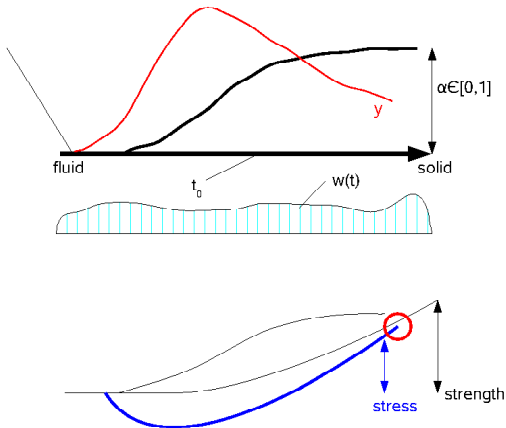
## controls

cast mixture  
 $q_1$  cement  
 $q_2$  fly ash  
 $q_3$  water  
 $q_4$  aggregate

initial  
 temperature  
 $y(0) = q_5$

stripping point  
 $t_0 = q_6$

water cooling  
 $w(t) = q_7(t)$



given:  $q \in \mathbb{R}^k$   
 find:  $u = (y, \tau)$  s.t.

$\Omega \subset \mathbb{R}^3$  polygonal  
 $y : \Omega \times [0, T] \rightarrow \mathbb{R}$  temperature  
 $\tau : \Omega \times [0, T] \rightarrow \mathbb{R}$  maturity

$$\left\{ \begin{array}{ll} \tau_t = g(y) & \text{in } \Omega \times (0, T) \\ \tau(0) = 0 & \text{in } \Omega \\ c(q)\rho(q)y_t - \lambda(q)\Delta y = Q_\infty(q)g(y)h(\tau, q) & \text{in } \Omega \times (0, T) \\ \lambda(q)\frac{\partial}{\partial n}y = \sigma(q)(\bar{y} - y) & \text{on } \Gamma \times (0, T) \\ y(0) = y_0(q) & \text{in } \Omega \end{array} \right.$$

short:  $u = S(q)$

$g, h, \dots$  model functions

given:  $q \in L^2(0, T)$   
 find:  $u = (y, \tau)$  s.t.

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$$\left\{ \begin{array}{ll} \tau_t = g(y) & \text{in } \Omega \times (0, T) \\ \tau(0) = 0 & \text{in } \Omega \\ c\rho y_t - \lambda \Delta y = Q_\infty g(y) h(\tau) - \dot{Q}_{pipe}(q(t)) & \text{in } \Omega \times (0, T) \\ \lambda \frac{\partial}{\partial n} y = \sigma(\bar{y} - y) & \text{on } \Gamma \times (0, T) \\ y(0) = y_0 & \text{in } \Omega \end{array} \right.$$

short:  $u = S(q)$

$g, h, \dots$  model functions

- theory of single aspects well understood
- macroscopic approach with interpolated model functions for physical quantities  $\Rightarrow$  model for simulation
  
- no optimization in literal sense yet  
 $\Rightarrow$  formulation as parabolic optimal control problem with state constraints
- challenges: choice of discretisation, complex dynamic, nonlinear optimization problems with state constraints



$$\begin{aligned} \min J(q, u) \\ u = S(q) \end{aligned}$$

- cost functional may contain
  - material / control costs  $J_1(q)$
  - quality term  $J_2(u)$
  - penalty costs for unwanted configurations

$$\begin{aligned} \min J(q, u) \\ u = S(q) \\ f(u) \geq 0 \end{aligned}$$

- cost functional may contain
  - material / control costs  $J_1(q)$
  - quality term  $J_2(u)$
  - penalty costs for unwanted configurations
  
- state constraint
  - temperature  $u \leq u_b$
  - stability  $f_{ct}(\tau) = f_{ct,\infty} \frac{\exp(a\tau^b) - \alpha_0}{1 - \alpha_0} \geq f_{ct,given}$
  - cracking models



$$(P) \begin{cases} \min J(q, u) \\ A(q, u)(\phi) = 0 \\ f(u) \geq 0 \end{cases} \quad \begin{array}{l} q \in L^2(I, R) \quad u \in W(0, T) \\ A.. \text{ weak formulation of parabolic pde} \\ \text{assume: optimality conditions hold} \\ \text{formulated via Lagrangian} \end{array}$$

$$\mathcal{L}(q, u, z, \mu) = J(q, u) - A(q, u)(z) + \langle f(u), \mu \rangle$$


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Regularization of the pointwise state constraint

$$\gamma \in \mathbb{R}_+, \quad \sigma > 1, \quad b_\gamma(u) := \int_Q \frac{1}{(\sigma-1)\gamma^\sigma} (f(u(t, x)))^{1-\sigma} d(t, x).$$

$$(P_\gamma) \begin{cases} \min J_\gamma(q_\gamma, u_\gamma) := J(q_\gamma, u_\gamma) + b_\gamma(u_\gamma) \\ A(q_\gamma, u_\gamma)(\phi) = 0 \end{cases}$$

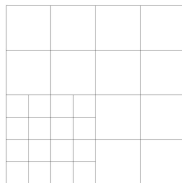
→ opt.cond. via

$$\mathcal{L}_\gamma(q_\gamma, u_\gamma, z_\gamma) = J(q_\gamma, u_\gamma) - A(q_\gamma, u_\gamma)(z_\gamma) + b_\gamma(u_\gamma)$$

discretization of  $(P_\gamma)$



implicit Euler in time,  $I = \cup I_m$



linear elements in space

- full discretization

$$X_{kh}^{01} = \{v_\gamma^{kh} \in L^2(L^2) : v_{kh}|_{I_m} \in \mathcal{P}_0(I_m, V_h^1) \forall m = 1..M\}$$

with  $V_h^1 = \{v_\gamma^h \in C(\bar{\Omega}) \cap H^1(\Omega) : v^h|_K \in \mathcal{Q}_1(K) \forall K \in \mathcal{T}_h\}$

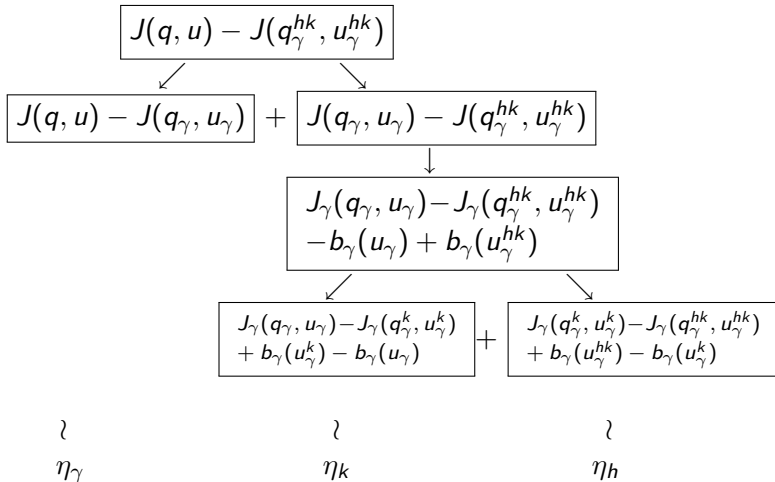
- solve discrete problems  $\rightarrow$  inexact Newton + LS-globalization on  $j(q) = J(q, S(q))$
- balance refinement of spatial, temporal meshes + increase  $\gamma \rightarrow \infty$   
 $\rightarrow$  adaptively, driven by a posteriori error estimation



$$J(q, u) - J(q_\gamma^{hk}, u_\gamma^{hk})$$

$$\begin{array}{c}
 \boxed{J(q, u) - J(q_\gamma^{hk}, u_\gamma^{hk})} \\
 \swarrow \quad \searrow \\
 \boxed{J(q, u) - J(q_\gamma, u_\gamma)} + \boxed{J(q_\gamma, u_\gamma) - J(q_\gamma^{hk}, u_\gamma^{hk})}
 \end{array}$$

$$\begin{array}{c}
 \boxed{J(q, u) - J(q_\gamma^{hk}, u_\gamma^{hk})} \\
 \swarrow \quad \searrow \\
 \boxed{J(q, u) - J(q_\gamma, u_\gamma)} + \boxed{J(q_\gamma, u_\gamma) - J(q_\gamma^{hk}, u_\gamma^{hk})} \\
 \downarrow \\
 \boxed{J_\gamma(q_\gamma, u_\gamma) - J_\gamma(q_\gamma^{hk}, u_\gamma^{hk})} \\
 \quad - b_\gamma(u_\gamma) + b_\gamma(u_\gamma^{hk})
 \end{array}$$



Let  $\xi = (q, u, z, \mu)$  be  $(P)$ -optimal,  $\hat{\xi}_\gamma = (q_\gamma, u_\gamma, z_\gamma)$  be  $(P_\gamma)$ -optimal, set  $\xi_\gamma = (q_\gamma, u_\gamma, z_\gamma, b'_\gamma(u_\gamma))$ .

## error representation

$$\begin{aligned}
 J(q, u) - J(q_\gamma, u_\gamma) &= \mathcal{L}(\xi) - \mathcal{L}(\xi) + \langle f(u_\gamma), b'_\gamma(u_\gamma) \rangle \\
 &= \frac{1}{2} \mathcal{L}'(\xi)(\xi - \xi_\gamma) + \frac{1}{2} \mathcal{L}'(\xi_\gamma)(\xi - \xi_\gamma) \\
 &\quad + \langle f(u_\gamma), b'_\gamma(u_\gamma) \rangle + \mathcal{R} \\
 &= \frac{1}{2} \langle f(u_\gamma) - f(u), b'_\gamma(u_\gamma) \rangle + \frac{1}{2} \langle f(u_\gamma), \mu \rangle + \mathcal{R}
 \end{aligned}$$

computable error estimator, heuristics



Let  $\xi = (q, u, z, \mu)$  be  $(P)$ -optimal,  $\hat{\xi}_\gamma = (q_\gamma, u_\gamma, z_\gamma)$  be  $(P_\gamma)$ -optimal, set  $\xi_\gamma = (q_\gamma, u_\gamma, z_\gamma, b'_\gamma(u_\gamma))$ .

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 &\quad + \langle f(u_\gamma), b'_\gamma(u_\gamma) \rangle + \mathcal{R} \\
 &= \frac{1}{2} \langle f(u_\gamma) - f(u), b'_\gamma(u_\gamma) \rangle + \frac{1}{2} \langle f(u_\gamma), \mu \rangle + \mathcal{R}
 \end{aligned}$$

computable error estimator, heuristics

$$\rightarrow \boxed{\eta_\gamma := c_0 \langle f(u_\gamma^{hk}), b'_\gamma(u_\gamma^{hk}) \rangle} \quad c_0 \in [0.5, 1]$$

depending on knowledge about  $f(u)$



- techniques from error estimation for unconstrained problems

$$\eta_k := \frac{1}{2}(\tilde{\rho}^u(\cdot)(I_k^{(1)} z_\gamma^{hk} - z_\gamma^{hk}) + \tilde{\rho}^z(\cdot)(I_k^{(1)} u_\gamma^{hk} - u_\gamma^{hk})) \\ - b'_\gamma(\cdot)(I_k^{(1)} u_\gamma^{hk} - u_\gamma^{hk})$$

$$\eta_h := \frac{1}{2}(\tilde{\rho}^u(\cdot)(I_{2h}^{(2)} z_\gamma^{kh} - z_\gamma^{kh}) + \tilde{\rho}^z(\cdot)(I_{2h}^{(2)} u_\gamma^{kh} - u_\gamma^{kh})) \\ - b'_\gamma(\cdot)(I_{2h}^{(2)} u_\gamma^{kh} - u_\gamma^{kh})$$

- techniques from error estimation for unconstrained problems

$$\eta_k := \frac{1}{2}(\tilde{\rho}^u(\cdot)(I_k^{(1)} z_\gamma^{hk} - z_\gamma^{hk}) + \tilde{\rho}^z(\cdot)(I_k^{(1)} u_\gamma^{hk} - u_\gamma^{hk})) - b'_\gamma(\cdot)(I_k^{(1)} u_\gamma^{hk} - u_\gamma^{hk})$$

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- error equilibration strategy

relative contributions of  $\eta_{abs} := |\eta_h| + |\eta_k| + |\eta_\gamma|$

$\frac{|\eta_h|}{\eta_{abs}} > c_1$

refine spatial mesh

$\frac{|\eta_k|}{\eta_{abs}} > c_1$

refine temporal  
mesh

$\frac{|\eta_\gamma|}{\eta_{abs}} > c_1$

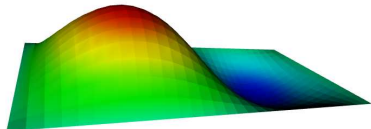
increase  $\gamma$   
 $\gamma_{new} := c_2 \gamma$

e.g.  $c_1 = 0.2, c_2 = 2$

$$\min \frac{1}{2} \|u - \sin(2\pi t) \sin(\pi x) \sin(\pi y)\|_{L^2(I \times \Omega)}^2$$

$$\begin{cases} u_t - \Delta u = 2 \sin(\pi x) \sin(\pi y) q(t) & \text{in } I \times \Omega \\ u(t, x) = 0 & \text{on } I \times \Gamma \\ u(0, x) = 0 & \text{on } \Omega \end{cases}$$

$$u(t, x) \leq 0.1 \text{ in } I \times \Omega$$



$$\Omega = [0, 2] \times [0, 1]$$

$$I = [0, 1]$$

$J(q^{hk}, u^{hk}) - J(q_\gamma^{hk}, u_\gamma^{hk})$	$\eta_\gamma$	$I_{eff, \gamma}$
-3.43e-03	-6.31e-03	0.543
-4.56e-04	-7.52e-04	0.607
-8.01e-05	-1.10e-04	0.728
-1.89e-05	-2.19e-05	0.863
-5.33e-06	-5.63e-06	0.947
-1.62e-06	-1.65e-06	0.982
-5.06e-07	-5.09e-07	0.994
-1.60e-07	-1.60e-07	0.998
-5.04e-08	-5.05e-08	0.999
-1.60e-08	-1.60e-08	1.000
-5.05e-09	-5.05e-09	1.000

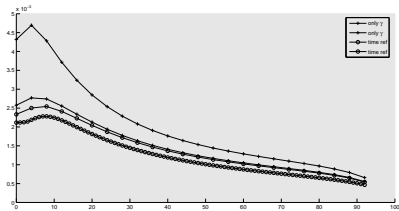
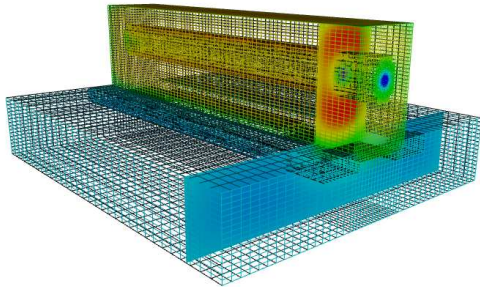
$$\begin{aligned}
 & \min \frac{1}{2} \|u - \sin(2\pi t) \sin(\pi x) \sin(\pi y)\|_{L^2(I \times \Omega)}^2 \\
 & \begin{cases} u_t - \Delta u = 2 \sin(\pi x) \sin(\pi y) q(t) & \text{in } I \times \Omega \\ u(t, x) = 0 & \text{on } I \times \Gamma \\ u(0, x) = 0 & \text{on } \Omega \end{cases} \\
 & u(t, x) \leq 0.1 \text{ in } I \times \Omega
 \end{aligned}$$

$N$	$M$	$\gamma$	$\eta_h$	$\eta_k$	$\eta_\gamma$	$J(q, u)^* - J_\gamma(q_\gamma^{hk}, u_\gamma^{hk})$	$l_{eff}$
45	6	3.16	2.50e-02	2.55e-06	-4.65e+00	-2.34e+00	0.507
45	6	10	2.56e-02	-7.12e-04	-4.71e-01	-2.52e-01	0.538
45	6	31.6	-7.53e-05	-2.19e-03	-5.49e-02	-3.30e-02	0.578
45	6	100	-1.31e-03	-2.43e-03	-6.35e-03	-6.91e-03	0.685
45	12	316	-1.56e-03	-9.21e-04	-8.08e-04	-3.36e-03	1.022
153	24	316	-3.90e-04	-3.63e-04	-8.83e-04	-1.39e-03	0.848
561	48	1000	-1.06e-04	-1.38e-04	-1.17e-04	-3.43e-04	0.949
2145	96	3160	-7.96e-05	-5.33e-05	-1.34e-05	-1.46e-04	1.000

water cooling

$$\text{find } q \in Q = L^2(0, T)$$

$$\text{s.t. } u \leq u_b$$



$$I_{eff, \gamma} = 0.731, 0.761, 1.011, \dots$$

optimal control problem: parabolic + state constraints

- solution via regularization of constraints
- a-posteriori estimation of error contributions
- equilibration strategy for  $\eta_h, \eta_k, \eta_\gamma$

mechanical properties of young concrete

- formulation as OCP in above problem class
- first optimization results
- cooperation with:
  - T. Apel, T. Flaig (UniBW Munich)
  - R. Nothnagel (ibmb Braunschweig)

**Thank you**

**for your attention!**