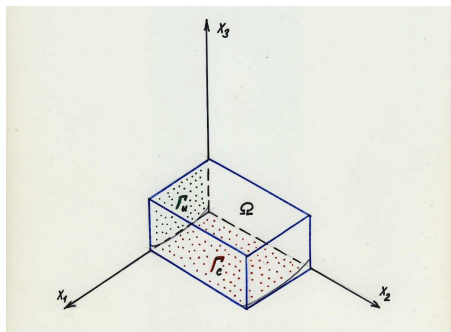


# Shape optimization in 3D contact problems with Coulomb friction

*Shape optimization in 3D contact problems  
with Coulomb friction*

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# Introduction



Elastic body:  $\Omega \subset \mathbb{R}^3$ ,  $bd\Omega = \Gamma_u \cup \Gamma_p \cup \Gamma_c$

$\Gamma_u$  ... zero displacements

$\Gamma_p$  ... surface tractions

$\Gamma_c$  ... boundary in contact with the "rigid" plane  
 $x_3 \times x_2$

After a FE discretization one obtains a finite-dimensional **Mathematical Program with Equilibrium Constraints (MPEC)**

$$\begin{array}{ll} \text{minimize} & f(\alpha, y) \\ \text{subject to} & y \in S(\alpha) \\ & \alpha \in U_{ad}. \end{array} \quad (1)$$

In (1) the **design variable**  $\alpha \in \mathbb{R}^p$  specifies the shape of the contact boundary and the **state variable**  $y = (u_\tau, u_\nu, \lambda) \in \mathbb{R}^{2p} \times \mathbb{R}^p \times \mathbb{R}_+^p$ , where

$p$  ..... number of nodes in the contact boundary

$u_\tau$  ..... vector of tangential displacements

$u_\nu$  ..... vector of normal displacements

$\lambda$  ..... multiplier associated with the **nonpenetrability** constraint

$$u_\nu + \alpha \geq 0.$$

Further, in (1)  $f$  is an objective,  $U_{ad}$  is the set of admissible design variables (controls) and the

control-state map  $S$  is given by the generalized equation (GE)

$$\begin{aligned}0 &\in A_{\tau\tau}(\alpha)u_{\tau} + A_{\tau y}(\alpha)u_y - l_{\tau}(\alpha) + \tilde{Q}(u_{\tau}, \lambda) \\0 &= A_{y\tau}(\alpha)u_{\tau} + A_{yy}(\alpha)u_y - l_y(\alpha) \\0 &\in u_y + \alpha + N_{\mathbb{R}_+^p}(\lambda),\end{aligned}\tag{2}$$

where  $A_{\tau\tau}, A_{\tau\nu}, A_{\nu\tau}, A_{\nu\nu}$  are blocks of appropriate restriction of stiffness matrix,  
 $b_\tau, b_\nu$  reflect the action of external forces

and

$$\tilde{Q}(u_\tau, \lambda) = F \begin{bmatrix} \lambda^1 \partial \|u_\tau^1\| \\ \lambda^2 \partial \|u_\tau^2\| \\ \vdots \\ \lambda^p \partial \|u_\tau^p\| \end{bmatrix}$$

with  $F > 0$  being the friction coefficient.

## Outline:

- (i) Implicit Programming Approach (IMP)
- (ii) Solution of the state problem
- (iii) Computation of subgradients (Sensitivity analysis)
- (iv) Test examples

# Implicit Programming Approach

Assume that

- $f$  is continuously differentiable;
- $U_{ad}$  is closed, and
- $F$  is sufficiently small.

Then  $S$  is single-valued and locally Lipschitz and (1) amounts to the mathematical program

$$\begin{aligned} & \text{minimize } \Theta(\alpha) (:= f(\alpha, S(\alpha))) \\ & \text{subject to } \alpha \in U_{ad} \end{aligned} \quad (3)$$

in variable  $\alpha$  only.



(3) can be solved by a bundle method of NDO provided  $\Theta$  is weakly semismooth.

Further, for each  $\bar{\alpha} \in U_{ad}$  we must be able

- to compute  $\bar{y} = S(\bar{\alpha})$ , i.e., to solve a 3D contact problem with Coulomb friction;
- to compute one arbitrary Clarke subgradient  $\xi \in \bar{\partial}\Theta(\bar{\alpha})$ .

# Solution of the state problem

Let us replace the multiplier  $\lambda$  in  $\tilde{Q}$  by a fixed **slip bound**  $g \in \mathbb{R}_+^p$ . Then (2) amounts to the optimization problem

$$\begin{aligned} & \text{minimize } \frac{1}{2} \langle u, A(\alpha) u \rangle - \langle \ell(\alpha), u \rangle + \sum_{i=1}^p F g^i \|u_{\tau}^i\| \\ & \text{subject to} \\ & \quad -u_{\nu} \leq \alpha. \end{aligned} \tag{4}$$

We introduce a new variable  $v \in \mathbb{R}^{2p}$  and put

$$u = \begin{bmatrix} v \\ \lambda \end{bmatrix}.$$

Then a dual problem to (4) attains the form

$$\begin{aligned} & \text{minimize } \frac{1}{2} \langle u, Fu \rangle - \langle h, u \rangle \\ & \text{subject to } \|v^i\| \leq Fg^i, \quad \lambda \geq 0, \end{aligned} \quad (5)$$

where  $F$  and  $h$  can be expressed in terms of  $A(\alpha)$ ,  $l(\alpha)$  and  $\alpha$ .

Method of successive approximations:

- 1° Initialize  $g_0$ , put  $k := 0$ ;
- 2° Solve (5) with  $g = g_k$  and compute thus  $u_k = (v_k, \lambda_k)$
- 3° Put  $g_{k+1} = \lambda_k$ ,  $k = k+1$  and go to 2°.

# Computation of subgradients (sensitivity analysis)

In most MPECs  $S$  is a  $PC^1$  mapping, i.e.,  $S$  is continuous and for all  $\alpha$

$$S(\alpha) \in \{S_1(\alpha), \dots, S_q(\alpha)\},$$

where  $S_1, \dots, S_q$  are continuously differentiable.

In such a case one has for the **Clarke** generalized Jacobian  $\bar{\partial}S(\bar{\alpha})$  the inclusion

$$\bar{\partial}S(\bar{\alpha}) \subset \text{conv} \{ \nabla S_i(\bar{\alpha}) \mid i \in I(\bar{\alpha}) \},$$

where

$$I(\bar{\alpha}) := \{ i \in \{1, 2, \dots, q\} \mid S(\bar{\alpha}) = S_i(\bar{\alpha}) \}$$

is the index set of **active pieces**.

Consequently,  
on the basis of the Clarke generalized Jacobian Chain Rule, one may put

$$\xi = \nabla_{\alpha} f(\bar{\alpha}, \bar{y}) + (\nabla S_i(\alpha))^T \nabla_y f(\bar{\alpha}, \bar{y}) \quad \text{for some } i \in I(\bar{\alpha}).$$

Unfortunately, in case of 3D contact problem with Coulomb friction we do not dispose with a representation of  $S$  showing that it is a  $PC^1$  mapping.

# Computation of subgradients (sensitivity analysis)

One could attempt to compute a matrix from  $\bar{\partial}S(\bar{\alpha})$  according to the definition as

$$\lim_{\alpha_i \rightarrow \bar{\alpha}} \{ \nabla S(\alpha_i) \mid S \text{ is differentiable at } \alpha_i \},$$

but it is not easy to identify all situations, when  $S$  is differentiable.

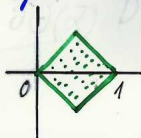
It holds that for all  $y^*$

$$(\bar{\partial}S(\bar{\alpha}))^T y^* = \text{conv } D^*S(\bar{\alpha}, \bar{y})(y^*)$$

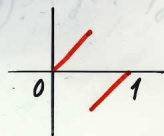
and so it suffices a vector from  $D^*S(\bar{\alpha})(\nabla_y f(\bar{\alpha}, \bar{y}))$ .

# Computation of subgradients (sensitivity analysis)

Example:  $S(\cdot) = \text{Proj}_{X_2}(\cdot)$  at its vertex,  $y^* = (1, 0)^T$ .



$\bar{\partial}S(0)y^*$



$D^*S(0)(y^*)$

Computation of  $D^*S(\bar{\alpha})$ :

(2) has the form

$$0 \in F(\alpha, y) + Q(y)$$

which yields

$$S(\alpha) = \left\{ y \mid \begin{bmatrix} y \\ -F(\alpha, y) \end{bmatrix} \in \text{gr} Q \right\}.$$

# Computation of subgradients (sensitivity analysis)

**Theorem 1.** If  $F$  is sufficiently small, then (2) satisfies the **strong regularity condition** at all points  $(\bar{\alpha}, \bar{y})$ , where  $\bar{\alpha} \in U_{ad}$  and  $\bar{y} = S(\bar{\alpha})$ . Consequently, for all  $y^*$

$$D^*S(\bar{\alpha})(y^*) \subset \left\{ (\nabla_{\alpha} F(\bar{\alpha}, \bar{y}))^T v \mid 0 \in y^* + (\nabla_y F(\bar{\alpha}, \bar{y}))^T v + D^*Q(\bar{y}, -F(\bar{\alpha}, \bar{y}))(v) \right\}$$

Equality holds provided  $gr Q$  is **normally reg.** at  $(\bar{y}, -F(\bar{\alpha}, \bar{y}))$ .



# Computation of subgradients (sensitivity analysis)

For the purpose of computation of  $D^*Q$  we will reorder  $y$  as  $(y^1, y^2, \dots, y^p) \in (\mathbb{R}^4)^p$  with

$$y^i = (u_x^i, u_y^i, \lambda^i) \in \mathbb{R}^2 \times \mathbb{R} \times \mathbb{R}_+, \quad i=1, 2, \dots, p$$

and reorganize accordingly the lines in (2). The multi-valued part attains then the form

$$Q(y) = \begin{bmatrix} \Phi(y^1) \\ \Phi(y^2) \\ \vdots \\ \Phi(y^p) \end{bmatrix} \quad \text{with} \quad \Phi(y^i) = \begin{bmatrix} F \lambda^i \partial \|u_x^i\| \\ 0 \\ N_{\mathbb{R}_+}(\lambda^i) \end{bmatrix}.$$

# Computation of subgradients (sensitivity analysis)

Consequently, with  $b = (b^1, b^2, \dots, b^p) \in Q(y)$  and  $b^* = (b^{*1}, b^{*2}, \dots, b^{*p})$ ,  $y^* = (y^{*1}, y^{*2}, \dots, y^{*p})$  one has

$$y^* \in D^*Q(y, b)(b^*) \iff y^{*i} \in D^*\Phi(y^i, b^i)(b^{*i})$$

for  $i = 1, 2, \dots, p$ .

So, it suffices to consider the  $i$ th node and investigate the mapping

$$\begin{bmatrix} b_1^i \\ b_2^i \end{bmatrix} \in \mathcal{F} \lambda^i \partial \|u_c^i\|$$

$$b_3^i = 0$$

$$b_4^i \in N_{\mathbb{R}_+}(\lambda^i).$$

# Computation of subgradients (sensitivity analysis)

We assume for simplicity that  $\mathcal{F} = 1$ , omit the index  $i$  and put  $b_{i2} = (b_1, b_2)^T$ . The position of  $(y, b)$  in  $\text{Gr}\Phi$  is specified in the following table:

	strong contact $\lambda > 0, b_3 = 0$	weak contact $\lambda = 0, b_3 = 0$	no contact $\lambda = 0, b_3 < 0$
sliding $u_T \neq 0$	$M_1$	$M_2$	L
weak sticking $u_T = 0$ $\ b_{i2}\  = \lambda$	$M_3^-$	$M_4$	
strong sticking $u_T = 0$ $\ b_{i2}\  < \lambda$	$M_3^+$		

# Computation of subgradients (sensitivity analysis)

Simple cases ( $M_1, L, M_3^+$ ):

$M_1$ :  $\exists$  a neighborhood  $\mathcal{U}$  of the reference point  $(\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda}, \bar{b}_1, \bar{b}_2, 0, \bar{b}_3) \in \text{Gr } \Phi$  such that

$$u_\tau \neq 0, \lambda > 0, b_{12} = \frac{u_\tau}{\|u_\tau\|}, b_3 = 0 \text{ on } \text{Gr } \Phi \cap \mathcal{U}.$$

consequently,

$$\left( \nabla \Phi(\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda}, \bar{b}) \right)^* = \begin{bmatrix} \bar{\lambda} \frac{(\bar{u}_{\tau 2})^2}{\|\bar{u}_\tau\|^3} & -\bar{\lambda} \frac{\bar{u}_{\tau 1} \bar{u}_{\tau 2}}{\|\bar{u}_\tau\|^3} & 0 & 0 \\ -\bar{\lambda} \frac{\bar{u}_{\tau 1} \bar{u}_{\tau 2}}{\|\bar{u}_\tau\|^3} & \bar{\lambda} \frac{(\bar{u}_{\tau 1})^2}{\|\bar{u}_\tau\|^3} & 0 & 0 \\ 0 & 0 & 0 & 0 \\ \frac{\bar{u}_{\tau 1}}{\|\bar{u}_\tau\|} & \frac{\bar{u}_{\tau 2}}{\|\bar{u}_\tau\|} & 0 & 0 \end{bmatrix}.$$

# Computation of subgradients (sensitivity analysis)

(L)  $\exists$  a neighborhood  $\mathcal{U}$  of the reference point  $(\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda}, \bar{b}_1, \bar{b}_2, 0, \bar{b}_4) \in \text{gr}\Phi$  such that

$$\lambda = 0, \quad b_{12} = 0, \quad b_4 < 0 \quad \text{on } \text{gr}\Phi \cap \mathcal{U}.$$

It follows that

$$N_{\text{gr}\Phi}(\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda}, \bar{b}) = \left\{ (\xi, \eta) \in \mathbb{R}^4 \times \mathbb{R}^4 \mid \xi_1 = \xi_2 = \xi_3 = 0, \right. \\ \left. \eta_4 = 0 \right\}.$$

Hence,

$$D^*\Phi(\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda}, \bar{b})(\eta) = \begin{cases} \{0\} \times \{0\} \times \{0\} \times \mathbb{R} & \text{if } \eta_4 = 0 \\ \emptyset & \text{otherwise.} \end{cases}$$

# Computation of subgradients (sensitivity analysis)

$M_3^+$   $\exists$  a neighborhood  $\mathcal{U}$  of the reference point such that

$$u_\tau = 0, \lambda > 0, \|b_{12}\| < \lambda, \quad b_4 = 0 \text{ on } \text{Gr}\Phi \cap \mathcal{U}.$$

It follows that

$$N_{\text{Gr}\Phi}(\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda}, \bar{b}) = \{(\xi, \eta) \in \mathbb{R}^4 \times \mathbb{R}^4 \mid \xi_3 = \xi_4 = 0, \eta_1 = \eta_2 = 0\}$$

Hence,

$$D^*\Phi(\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda}, \bar{b})(\eta) = \begin{cases} \mathbb{R} \times \mathbb{R} \times \{0\} \times \{0\} & \text{if } \eta_1 = \eta_2 = 0 \\ \emptyset & \text{otherwise.} \end{cases}$$

# Computation of subgradients (sensitivity analysis)

$M_2$  (weak contact, sliding)

**Proposition 2.** Let

$$F(x, y, z) = \begin{bmatrix} G(x, y) \\ H(y, z) \end{bmatrix},$$

where  $G, H$  are closed-graph multifunctions.  
Assume that

$$\begin{bmatrix} \bar{d}_1 \\ \bar{d}_2 \end{bmatrix} \in F(\bar{x}, \bar{y}, \bar{z}),$$

and the qualification condition

$$\begin{bmatrix} 0 \\ w_2 \end{bmatrix} \in DG^*(\bar{x}, \bar{y})(0) \ \& \ \begin{bmatrix} -w_2 \\ 0 \end{bmatrix} \in DH^*(\bar{y}, \bar{z})(0) \Rightarrow w_2 = 0$$

is fulfilled.

# Computation of subgradients (sensitivity analysis)

Then one has

$$D^*F(\bar{x}, \bar{y}, \bar{z})(d_1^*, d_2^*) \subset \left\{ (w_1, w_2 + w_3, w_4) \mid \right. \\ \left. (w_1, w_2) \in D^*G(\bar{x}, \bar{y})(d_1^*), (w_3, w_4) \in D^*H(\bar{y}, \bar{z})(d_2^*) \right\}$$

If  $G$  is continuously differentiable, then

$$D^*F(\bar{x}, \bar{y}, \bar{z})(d_1^*, d_2^*) =$$

$$\left\{ (\nabla_x G(\bar{x}, \bar{y}))^T d_1^*, (\nabla_y G(\bar{x}, \bar{y}))^T d_1^* + w_3, w_4 \mid (w_3, w_4) \in D^*H(\bar{y}, \bar{z})(d_2^*) \right\}$$



# Computation of subgradients (sensitivity analysis)

**Proposition 3.** Let the reference point  $(\bar{y}, \bar{b})$ ,  $\bar{y} = (\bar{u}_\tau, \bar{u}_\nu, \bar{\lambda})$  belong to  $M_2$ . Then one has

$$D^*\Phi(\bar{y}, \bar{b})(\eta) = \left\{ \begin{array}{l} 0 \\ 0 \\ 0 \\ \frac{\bar{y}_1}{\|\bar{y}_{12}\|} \eta_1 + \frac{\bar{y}_2}{\|\bar{y}_{12}\|} \eta_2 + w \end{array} \right\} :$$

$w \in \left\{ \begin{array}{l} \mathbb{R} \text{ provided } \eta_3 = 0 \\ \mathbb{R}_- \text{ provided } \eta_3 < 0 \\ 0 \text{ otherwise.} \end{array} \right\}$  approaching via  $L$   
 $M_2$   
 $M_1$

*Remark.* By approaching  $(\bar{y}, \bar{b})$  via  $L$  we delete the contact condition which has no influence. This leads to the simplest adjoint GE and therefore this option has been used in the computer code. By approaching  $(\bar{y}, \bar{b})$  via  $M_1$ , the contact condition amounts to an equality.

# Computation of subgradients (sensitivity analysis)

$M_3^-$  (strong contact, weak sticking)

**Proposition 4.** Consider a point  $(\bar{y}, \bar{b})$  from  $M_3^-$  and put  $w = \frac{\bar{b}_{12}}{\lambda} (\in S_2)$ . Then one has

$$N_{\text{gr}\Phi}(\bar{y}, \bar{b}) = \left\{ (\xi, \eta) \in \mathbb{R}^4 \times \mathbb{R}^4 \mid \xi_{12} = 0, \xi_3 = 0, (\eta_{12}, \xi_4) \in \mathbb{R}^{(w, -1)} \right\}$$

$$M_4 \leftarrow \cup \left\{ (\xi, \eta) \in \mathbb{R}^4 \times \mathbb{R}^4 \mid \xi_{12} \in w^\perp, \xi_3 = 0, \xi_4 = 0, \eta_{12} = 0 \right\}$$

$$M_3^+ \cup \left\{ (\xi, \eta) \in \mathbb{R}^4 \times \mathbb{R}^4 \mid \xi_3 = 0, \xi_4 = 0, \eta_{12} = 0 \right\}$$

$$M_3^- \cup \left\{ (\xi, \eta) \in \mathbb{R}^4 \times \mathbb{R}^4 \mid \langle \xi_{12}, w \rangle \leq 0, \xi_3 = 0, (\eta_{12}, \xi_4) \in \mathbb{R}_+^{(w, -1)} \right\}$$

# Computation of subgradients (sensitivity analysis)

*Remark.* The first two sets exhaust accumulation points of sequences  $\xi^{(i)}, \eta^{(i)}$  such that

$$\xi^{(i)} = (\nabla \Phi(u_c^{(i)}, u_d^{(i)}, \lambda^{(i)}))^T \eta^{(i)},$$

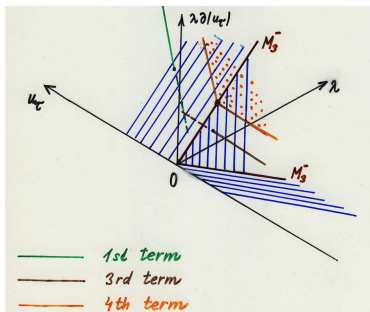
provided  $u_c^{(i)} \rightarrow 0$  and  $\lambda^{(i)} \rightarrow \bar{\lambda} > 0$ .

*Remark.* By approaching  $(\bar{y}, \bar{b})$  via  $M_3^+$  we do not have to do with  $w$  and get a simple formula. Hence, this option is used in the computer code.

# Computation of subgradients (sensitivity analysis)

Illustration of the above result in 2D,  
where

$$\Phi(u_x, u_y, \lambda) = \begin{bmatrix} \lambda \partial |u_x| \\ 0 \\ N_{R_x}(\lambda) \end{bmatrix}$$



The 2nd term does not have an analogue in the 2D (polyhedral) case.

$$f(\alpha, y) = \|\lambda - \bar{\lambda}_y\|^2,$$

where  $\bar{\lambda}_y$  is a "desired" vector of normal stresses on the contact boundary.

The control vector  $\alpha$  represents control points of Bezier surface.

$U_{ad}$  comprises the constraint

$$\text{vol}(\Omega(\alpha)) = \text{const.}$$

and constraints restricting the slopes of  $\Gamma_c$ .

Discretization :  $25 \times 12 \times 12$  (3600 nodes)  
 $p = 300$  (1200 state variables)  
 $\dim \alpha = 32$

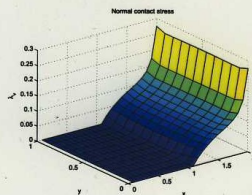
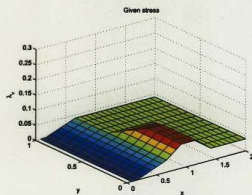
stopping criterion :  $\epsilon = 10^{-6}$

# Test examples

**Example:**

$$\min \|\bar{\lambda}_\nu - \lambda_\nu\|_2^2,$$

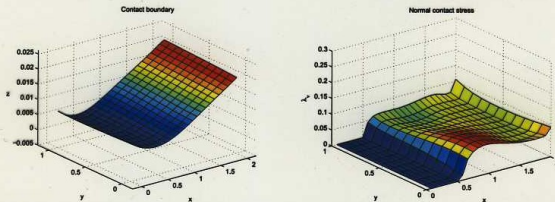
$$\varepsilon = 1 \cdot 10^{-4}$$



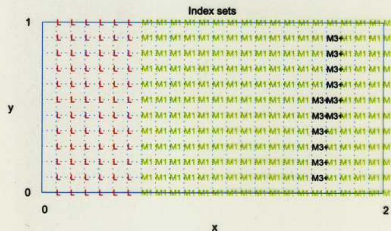
Target normal stress (left) and normal stress for initial design (right).



# Test examples



Optimal design (left) and normal stress for optimal design (right).



Index sets for optimal design.

## Open questions

- 1) Semismoothness of  $S_i$ ;
- 2)  $D^*\Phi$  in the case of  $M_i$  exactly.

## Further plans

- 1) To replace BT by a faster solver.
- 2) To rewrite all parts of the code to one language.