Nonconvex SDP Problems of Structural Optimization

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Outline

- Structural design with stability, vibration control
- FMO—a particular case of structural design
- Solving nonconvex SDP by PENNON
- Examples

MPEC:

 $\min_{oldsymbol{
ho},u}F(oldsymbol{
ho},u)$ s.t. $oldsymbol{
ho}\in U_{\mathrm{ad}}$ u solves $\mathcal{E}(oldsymbol{
ho},u)$

F(ho,u)		cost functional (weight, stiffness, peak stress)
ρ	• • •	design variable (thickness, material properties, shape)
$oldsymbol{u}$	•••	state variable (displacements, stresses)
$U_{ m ad}$	•••	admissible designs

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WEIGHT versus STIFFNESS:

- **9** W weight $\sum \rho_i$

Equilibrium constraint: u solves $\mathcal{E}(\rho, u) \longrightarrow \sum (\rho_i K_i) u = f$

WEIGHT versus STIFFNESS:

- **W** weight $\sum \rho_i$
- $f^T u$ C stiffness (compliance)

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S. Timoshenko:

Experience showed that structures like bridges or aircrafts may fail in some cases not on account of high stresses but owing to insufficient elastic stability.

Three quantities to control:

- **9** W weight $\sum \rho_i$
- λ min. eigenfrequency

 $K(\rho)u = \lambda M(\rho)u$

 $f^T u$

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- W weight $\sum \rho_i$
- C stiffness (compliance)
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- \checkmark C stiffness (compliance)
 - nce) $f^T u$
- Mathematical Mathematical Action Control Action C



Three quantities to control:

- **9** W weight $\sum \rho_i$
- λ critical buckling force $K(\rho)u = \lambda G(\rho, u)u$



Lowest (positive) eigenvalue of

$$K(
ho)u = {oldsymbol \lambda} G(
ho,u)u$$

(critical force) should be bigger than 1.

$$\min_{\substack{\rho, u}} W(\rho)$$

s.t.
$$K(\rho)u = f$$
$$f^T u \leq \widehat{C}$$
$$\rho_i \geq 0, \quad i = 1, \dots, m$$
$$\lambda \geq 1$$

Two standard tricks:

$$egin{aligned} K(oldsymbol{
ho}) &\succ 0, & u = K(oldsymbol{
ho})^{-1}f \ & f^T K(oldsymbol{
ho})^{-1}f \leq \widehat{C} & \Longleftrightarrow & igg(egin{aligned} \widehat{C} & f^T \ f & K(oldsymbol{
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ho})u &= oldsymbol{\lambda} G(oldsymbol{
ho},u)u \ &oldsymbol{\lambda} \geq 1 \end{aligned} ightarrow K(oldsymbol{
ho}) - G(oldsymbol{
ho},u) \succeq 0 \ & \Longleftrightarrow \quad K(oldsymbol{
ho}) - \widetilde{G}(oldsymbol{
ho}) \succeq 0 \ & \widetilde{G}(oldsymbol{
ho}) = G(oldsymbol{
ho},K(oldsymbol{
ho})^{-1}f) \end{aligned}$$

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Formulated as SDP problem:

 $egin{aligned} \min_{
ho} W(
ho) \ & ext{subject to} \ & K(
ho) - \widetilde{G}(
ho) \succeq 0 \ & igg(egin{aligned} c & f^T \ f & K(
ho) \end{pmatrix} \succeq 0 \ &
ho_i \geq 0, \quad i=1,\ldots,m \end{aligned}$

where

$$K(
ho) = \sum
ho_i K_i, \qquad \widetilde{G}(
ho) = \sum \widetilde{G}_i$$

Given an amount of material, boundary conditions and external load f, find the material (distribution) so that the body is as stiff as possible under f.

The design variables are the material properties at each point of the structure.

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$$\inf_{\substack{E \succcurlyeq 0 \\ \int tr(E) dx \leq 1}} \sup_{u \in U} \ -\frac{1}{2} \int_{\Omega} \langle \underline{E}e(u), e(u) \rangle \, dx + \int_{\Gamma_2} f \cdot u \, dx$$

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$$\inf_{\substack{\rho \ge 0 \\ \int \rho \, dx \le 1}} \sup_{u \in U} -\frac{1}{2} \int_{\Omega} \rho \langle e(u), e(u) \rangle \, dx + \int_{\Gamma_2} f \cdot u \, dx$$

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$$\inf_{lpha \in \mathbb{R}, u \in U} \left\{ lpha - f^T u \, | \, lpha \geq rac{m}{2} u^T A_i \, u \quad ext{for} \quad i = 1, \dots, m
ight\}$$

FMO, example



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FMO, example



Formulated as SDP problem:

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PENNON for SDP

 $egin{aligned} rac{ extsf{Problem:}}{x\in \mathbb{R}^n} & \min_{x\in \mathbb{R}^n} \left\{ b^Tx: \mathcal{A}(x) \preccurlyeq 0
ight\} \ \mathcal{A}: \mathbb{R}^n \longrightarrow \mathbb{S}_d \end{aligned}$

Notation:

 $\begin{array}{ll} \langle A,B\rangle_{\mathbb{S}_d} &:= {\rm tr}\left(A^TB\right) {\it inner product } {\rm on} \ \mathbb{S}_d\\ \mathbb{S}_{d_+} &= \{A\in \mathbb{S}_d \mid A {\it positive semidefinite}\}\\ U\in \mathbb{S}_{d_+} & {\it matrix multiplier (dual variable)}\\ \Phi_p & {\it penalty function } {\rm on} \ \mathbb{S}_d \end{array}$

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PENNON for SDP: algorithm

Generalized augmented Lagrangian algorithm for SDP: (based on modified barrier method of R. Polyak, 1992)

We have

$$\mathcal{A}(x) \preccurlyeq 0 \Longleftrightarrow \Phi_p(\mathcal{A}(x)) \preccurlyeq 0$$

and the corresponding *augmented Lagrangian*

$$F(x,U,p):=f(x)+\langle U,\Phi_p(\mathcal{A}(x))
angle_{\mathbb{S}_d}$$

Algorithm:

$$\begin{array}{ll} (i) & \mbox{Find } x^{k+1} \mbox{ satisfying } \| \nabla_{\!x} F(x,U^k,p^k) \| \leq \epsilon^k \\ (ii) & U^{k+1} = D_{\!\mathcal{A}} \, \Phi_p(\mathcal{A}(x);U^k) \\ (iii) & p^{k+1} < p^k \end{array}$$

Best choice of Φ : $\Phi(A) = (A - I)^{-1} - I$

PENNON for SDP: theory

Based on Breitfeld-Shanno, 1993; generalized by M. Stingl, 2003

Assume:

- 1. $f, \mathcal{A} \in C^2$
- 2. $x \in \Omega$ nonempty, bounded
- 3. Constraint Qualification

Then \exists an index set \mathcal{K} so that:

- ${\scriptstyle { { J } \hspace{-.05in} { J } \hspace{-.05in} } } U_k \to \widehat{U}, \ k \in {\cal K}$
- (\hat{x}, \hat{U}) satisfies first-order optimality conditions

PENNON for SDP: Hessian

The reciprocal barrier function in SDP

$$\Phi(A) = (A - I)^{-1} - I$$

Hessian

-

$$\begin{split} \frac{\partial^2}{\partial x_i \partial x_j} \Phi(\mathcal{A}(x)) &= \\ (\mathcal{A}(x) - I)^{-1} \frac{\partial \mathcal{A}(x)}{\partial x_i} (\mathcal{A}(x) - I)^{-1} \frac{\partial \mathcal{A}(x)}{\partial x_j} (\mathcal{A}(x) - I)^{-1} \\ &+ (\mathcal{A}(x) - I)^{-1} \frac{\partial^2 \mathcal{A}(x)}{\partial x_i \partial x_j} (\mathcal{A}(x) - I)^{-1} \\ &+ (\mathcal{A}(x) - I)^{-1} \frac{\partial \mathcal{A}(x)}{\partial x_j} (\mathcal{A}(x) - I)^{-1} \frac{\partial \mathcal{A}(x)}{\partial x_i} (\mathcal{A}(x) - I)^{-1} \end{split}$$

FMO with stability constraint (nonconvex SDP)

 $K(\mathbf{\rho}) + G(\mathbf{\rho}) \succeq 0$

$$K(oldsymbol{
ho}) = \sum_{e=1}^M oldsymbol{
ho}_e K_e$$

$$egin{aligned} G(oldsymbol{
ho}) &= \sum_{e=1}^M G_e & G_e(oldsymbol{
ho}) = \sum_{k=1}^K B_{e,k}^T S_{e,k}(oldsymbol{
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ho}) &= egin{pmatrix} \sigma_1 & \sigma_3 \ \sigma_3 & \sigma_2 \end{pmatrix} & \sigma_{e,k}(oldsymbol{
ho}) = oldsymbol{
ho}_e^{\widetilde{B}} \widetilde{B}_{e,k}^T (K^{-1}(oldsymbol{
ho}) f)_e \end{aligned}$$

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$$G(\rho) = \sum_{e=1}^{M} G_e \qquad G_e(\rho) = \sum_{k=1}^{K} B_{e,k}^T S_{e,k}(\rho) B_{e,k}$$
$$S_{e,k}(\rho) = \begin{pmatrix} \sigma_1 & \sigma_3 \end{pmatrix} \qquad \sigma_{e,k}(\rho) = \rho_e \widetilde{B}^T (K^{-1}(\rho)f)$$

$$S_{e,k}(\rho) = \begin{pmatrix} \sigma_1 & \sigma_3 \\ \sigma_3 & \sigma_2 \end{pmatrix} \qquad \sigma_{e,k}(\rho) = \rho_e B_{e,k}^T (K^{-1}(\rho)f)_e$$

memory: $O(M^2)$ (M = 500 \approx 64 MB)

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CPU: $O(K^2 * d^2 * M^3)$ for one Hessian assembling

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memory: $O(M^2)$ (M = 500 \approx 64 MB)

CPU: $O(K^2 * d^2 * M^3)$ for one Hessian assembling All dense matrix-matrix multiplications implemented in BLAS

FMO with stability constraint (nonconvex SDP)

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memory: $O(M^2)$ (M = 500 \approx 64 MB)

CPU: $O(K^2 * d^2 * M^3)$ for one Hessian assembling Pentium 4, 2.4GHz, ~100 Newton steps: 400 elements8h 45 min, 1000 elements~130 hours

FMO with vibration constraint (linear SDP)



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FMO with vibration constraint (linear SDP)



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FMO with vibration constraint (linear SDP)





Linear SDP, SDPA input file (Pentium 4, 2.5 GHz):

	no. of	size of
problem	variables	matrix
shmup-3	420	1801+840
shmup-4	800	3361+1600
shmup-5	1800	7441+3660

problem	PENNON	SDPT3	SDPA	DSDP	CSDP	SeDuMi
shmup-3	381	417	497	439	1395	23322
shmup-4	2095	2625	2952	2798	5768	>127320
shmup-5	14149	23535	m	fail	m	m

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FMO with vibration constraint (linear SDP)



FMO with stability constraint (nonlinear SDP)





FMO with vibration constraint (linear SDP)



shmup3 (420 elements) ... 6 min 20 sec

FMO with stability constraint (nonlinear SDP)





shmup3 (420 elements) ... 8 hours

FMO with vibration constraint (linear SDP)



shmup3 (420 elements) ... 6 min 20 sec

FMO with stability constraint (nonlinear SDP)



shmup3 (420 elements) ... 8 hours

shmup3 with no SDP constraints (convex NLP) ... 1 sec

Conclusions (so far)

PENNON algorithm works well for nonconvex SDP –accurate solution within 60–100 internal iterations– (more experience from BMI problems and truss design)

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complexity of second-order method too high (for "large" problems)

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FIRST-ORDER METHOD

Use conjugate gradient method for solving the Newton system

Use finite difference formula for Hessian-vector products:

$$\nabla^2 F(x_k) v \approx \frac{\nabla F(x_k + hv) - \nabla F(x_k)}{h}$$

with $h = (1 + \|x_k\|_2 \sqrt{arepsilon})$

Complexity: Hessian-vector product = gradient evaluation need for Hessian-vector-product type preconditioner

Limited accuracy (4–5 digits)

Nonlinear SDP—FMO with stability constraints

Can CG + approx. Hessian help?

Partly...

No preconditioning, approx. Hessian: as many gradient evaluations as CG steps (good) CG with no preconditioning inefficient (bad)

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Evaluation of exact diagonal as expensive as evaluation of full Hessian Evaluation of approx. diagonal

Only L-BFGS preconditioner can be used — but it isn't really efficient

Conclusions, part II

Hessian-free SDP:

First promising results, more testing (and coding) needed

Solving vibration problem as GEVP

Another option (vibration problems): Solve the maximum eigenvalue problem formulated as GEVP:

 λ min. eigenfrequency of $K(\rho)u = \lambda M(\rho)u$



Solving vibration problem as GEVP

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 λ min. eigenfrequency of $K(\rho)u = \lambda M(\rho)u$



(quasiconvex) SDP problem with BMI constraints — solve by PENBMI

Solving vibration problem as GEVP (example)

FMO with vibration constraint (linear SDP)



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Solving vibration problem as GEVP (example)

FMO with vibration constraint (linear SDP)



FMO with vibration constraint: BMI formulation)



THE END

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