FROM OPTIMALITY CRITERIA TO STRUCTURAL APPROXIMATIONS

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INTRODUCTION & MOTIVATION

Introduction

Structural optimization applied to sizing (weight minimization) problem

 $\begin{array}{ll} \min & W(x) = \sum_{i=1}^{n} w_{i} x_{i} \\ x \\ \text{s.t.:} & g_{j}(x) \leq 0 \qquad (j = 1 \dots m) \\ & \underline{x}_{i} \leq x_{i} \leq \overline{x}_{i} \qquad (i = 1 \dots n) \end{array}$

- Finite element model
- Design variables are the transverse sizes of the structural members (Fixed geometry and material properties)
- Design restrictions

$$g_j(x) \equiv u_j(x) - \overline{u}_j \leq 0$$

$$\sigma_j(x) - \overline{\sigma}_j \leq 0$$

$$\underline{\omega}_j^2 - \omega_j^2(x) \leq 0$$

$$\underline{\lambda}_j^2 - \lambda_j^2(x) \leq 0$$

Introduction

- Design constraints $g_i(x) < 0$
 - Implicit functions
 - Non linear functions
 - One constraint evaluation requires a complete FE analysis
- □ Side constraints: simple and explicit
 - Fabrication / technological / physical constraints
 - Treated separately in most methods
- □ Iterative process \rightarrow HIGH COST

INTRODUCTION

- Optimality criteria techniques (OC)
 - Highly specific
 - Intuitive techniques, simple
 - Convergence to a design that is not necessarily optimal (KKT conditions)
 - Difficulties in identifying the set of active constraints
 - Convergence instabilities
 - Small number of reanalyses, independent of the number of design variables

Résumé

- Low cost
- But uncertainty convergence

INTRODUCTION

- Pure Mathematical Programming methods
 - Very general
 - Rigorous methods, quite elaborated
 - Convergence to a local minimum
 - Stable and monotonic convergence
 - Large number of reanalyses, growing with the number of design variables

Résumé

- Rigorous framework & guaranteed convergence
- High cost (Growing with the size of the problem)

□ Theorem of virtual work,

$$T.V. = \tilde{\mathbf{q}}^T \mathbf{g} = \mathbf{q}^T \tilde{\mathbf{g}} = \tilde{\mathbf{q}}^T \mathbf{K} \tilde{\mathbf{q}}$$

Applying a virtual load vector (unit load vector) in the direction of under the displacement u:

$$\tilde{\mathbf{g}} = (0 \ 0 \ 0 \dots 1 \dots 0 \ 0)^T$$

□ The Principle of Virtual Work yiels

$$T.V. = \mathbf{q}^T \, \tilde{\mathbf{g}} = u \times 1 = \mathbf{q}^T \, \mathbf{K} \, \tilde{\mathbf{q}}$$

D With

$$\mathbf{K} = \sum_{i} \mathbf{L}_{i}^{T} \mathbf{K}_{i} \, \mathbf{L}_{i}$$

 For truss and plate design variables, the stiffness matrix takes the interesting form:

$$K_e = x_e \, \bar{K}_e$$

- Truss structures $x_e = A_e$
- Plate structures $x_e = t_e$
- One can decompose the contribution of each element:

$$u = \boldsymbol{q}^T \boldsymbol{K} \, \tilde{\boldsymbol{q}} = \sum_e \boldsymbol{q}_e^T \boldsymbol{K}_e \, \tilde{\boldsymbol{q}}_e$$
$$= \sum_e (x_e \, \boldsymbol{q}_e^T \bar{\boldsymbol{K}} \, \tilde{\boldsymbol{q}}_e)$$

The flexibility coefficients are constant for statically determinate structures

$$\boldsymbol{q}_e^T \boldsymbol{K}_e \; \tilde{\boldsymbol{q}}_e \; x_e = c_e$$

- For other structures, one can also assume a moderate redistribution of the internal loads around the current design point and has also constant value.
- Considering that the coefficients c_e are constant, Berke's criterion provides an explicit expression of the displacement u in terms of the design variables

$$\tilde{u} = \sum_{e} \frac{c_e^0}{x_e} \quad \text{with} \quad c_e^0 = (\boldsymbol{q}_e^{0T} \, \boldsymbol{K}_e^0 \, \tilde{\boldsymbol{q}}_e^0) \, x_e^0$$

BERKE'S APPROXIMATION ARE FIRST ORDER APPROXIMATIONS

A first order explicit approximation of displacement

 The Berke's expansion provide a first order explicit approximation of the displacement around x⁰

$$\tilde{u} = \sum_{i} \frac{c_i^0}{x_i}$$
 with $c_i^0 = (\boldsymbol{q}_i^{0T} \ \boldsymbol{K}_i^0 \ \tilde{\boldsymbol{q}}_i^0) \ x_i^0$

- In general (indeterminate structures) the c_i⁰ are not constant
- The expression is exact for statically determinate structures, but for statically indeterminate structures, it is only exact in the current point x⁰
- The Berke's expression is an approximation of the displacement u around the current design point x⁰ in terms of the design variables.

A first order explicit approximation of displacement

• The value of the approximation is exact in x_i⁰

$$\tilde{u}(\boldsymbol{x}_{i}^{0}) \ = \ \sum_{i} \frac{c_{i}^{0}}{x_{i}^{0}} \ = \ \sum_{i} \frac{(\boldsymbol{q}_{i}^{0T} \ \boldsymbol{K}_{i}^{0} \ \tilde{\boldsymbol{q}}_{i}^{0}) \ \boldsymbol{x}_{i}^{0}}{x_{i}^{0}} \ = \ \sum_{i} (\boldsymbol{q}_{i}^{0T} \ \boldsymbol{K}_{i}^{0} \ \tilde{\boldsymbol{q}}_{i}^{0}) \ = \ \boldsymbol{u}^{0}$$

- As c_i⁰ remains constant only along D(x⁰). It is also true for all points along the scaling line
- The derivatives of the approximations are exact in x_i⁰

$$\left. \frac{\partial \tilde{u}}{\partial x_i} \right|_{\boldsymbol{x}^0} = \left. \frac{\partial u}{\partial x_i} \right|_{\boldsymbol{x}^0}$$

APPROXIMATION IS A FIRST ORDER TAYLOR EXPANSION IN THE RECIPROCAL VARIABLE SPACE

Berke's approximation of the displacement *u*.

$$\tilde{u} = \sum_{i} \frac{c_i^0}{x_i}$$
 with $c_i^0 = (\mathbf{q}_i^{0T} \mathbf{K}_i^0 \tilde{\mathbf{q}}_i^0) x_i^0$

Suggest to use the reciprocal variables

$$z_i = \frac{1}{x_i}$$

• The Berke's approximation can take the simple form

$$\tilde{u}_j(\mathbf{z}) = \sum_i c_{ij} z_i$$

 We now show that the Berke's approximation is in fact the first order Taylor expansion of the displacement in the reciprocal space.

$$u(\mathbf{z}) = u(\mathbf{z}^0) + \sum_i \left(\frac{\partial u_j}{\partial z_i}\right)_0 (z_i - z_i^0) \leq \bar{u}$$

- In order to prove that one has to show that
 - C_{ij} are the derivative of u with respect to z_i:

$$c_i = \left. \frac{\partial u}{\partial z_i} \right|_{\mathbf{z}^0}$$

• The constant terms in '0' cancel each other

$$u(\mathbf{z}^0) = \sum_i c_i z_i^0$$

It comes

$$c_{i} = \frac{\partial u}{\partial z_{i}}\Big|_{\mathbf{z}^{0}} \qquad u(\mathbf{z}) = u(\mathbf{z}^{0}) + \sum_{i} \left(\frac{\partial u_{j}}{\partial z_{i}}\right)_{0} (z_{i} - z_{i}^{0}) \leq \bar{u}$$
$$u(\mathbf{z}) = u(\mathbf{z}^{0}) + \sum_{i} c_{i} (z_{i} - z_{i}^{0}) \leq \bar{u}$$
$$u(\mathbf{z}^{0}) = \sum_{i} c_{i} z_{i}^{0} \qquad u(\mathbf{z}) = [u_{j}(\mathbf{z}^{0}) - \sum_{i} c_{i} z_{i}^{0}] + \sum_{i} c_{i} z_{i} = \sum_{i} c_{i} z_{i}$$

Finally

$$\tilde{u} = \sum_{i} c_i z_i = \sum_{i} \frac{c_i}{x_i}$$

 Let's show that the virtual energy densities c_i are the first derivatives (gradients) of the constraints with respect to the reciprocal variables z_i=1/x_i that is

$$c_i = \left. \frac{\partial u}{\partial z_i} \right|_{\mathbf{z}^0}$$

• For the approximation obviously, we have :

$$\left. \frac{\partial \tilde{u}}{\partial z_i} \right|_{\mathbf{z}^0} = c_i = x_i^0 \, \mathbf{q}_i^{0T} \mathbf{K}_i \, \tilde{\mathbf{q}}_i^0$$

• Derivative of the (real) displacement with respect to the reciprocal design variable $z_i = 1/x_i$. It is clear that

$$\frac{\partial u}{\partial z_i} = \frac{\partial u}{\partial x_i} \frac{\partial x_i}{\partial z_i}$$

Since

$$\frac{\partial x_i}{\partial z_i} = \frac{\partial}{\partial z_i} \left(\frac{1}{z_i}\right) = -\frac{1}{z_i^2} = -x_i^2$$

It comes

$$\begin{aligned} \frac{\partial u}{\partial z_i} \Big|_{\mathbf{z}^0} &= \left. \frac{\partial u}{\partial x_i} \right|_{\mathbf{x}^0} (-x_i^{0\,2}) \\ &= \left. -\frac{\mathbf{q}_i^{0T} \mathbf{K}_i \tilde{\mathbf{q}}_i^0}{x_i^0} (-x_i^{0\,2}) \right. \\ &= \left. \mathbf{q}_i^{0T} \mathbf{K}_i \tilde{\mathbf{q}}_i^0 x_i^0 = c_i^0 = \left. \frac{\partial \tilde{u}}{\partial z_i} \right|_{\mathbf{z}^0} \end{aligned}$$

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We have to prove

$$u(\mathbf{z}^0) = \sum_i c_i z_i^0$$

□ This is performed by writing the expression of $u(\mathbf{z}_0)$

$$u(\mathbf{z}^{0}) = u(\mathbf{x}^{0}) = \sum_{i} \mathbf{q}_{i}^{0T} \mathbf{K}_{i} \, \tilde{\mathbf{q}}_{i}^{0} = \sum_{i} \mathbf{q}_{i}^{0T} \mathbf{K}_{i} \, \tilde{\mathbf{q}}_{i}^{0} \frac{x_{i}^{0}}{x_{i}^{0}} = \sum_{i} \frac{c_{i}^{0}}{x_{i}^{0}} = \sum_{i} c_{i}^{0} z_{i}^{0}$$

It comes that the approximation writes

$$u(\mathbf{z}) = [u_j(\mathbf{z}^0) - \sum_i c_i z_i^0] + \sum_i c_i z_i = \sum_i c_i z_i$$

First order explicit approximation of the stress constraints

- Previous interpretation of FSD and Berke's approximation suggests to generalize the first order approximation approach and to build the same high-quality approximations for stress and displacement constraints.
- For truss: stress constraints can be equivalent to relative displacements. But in general

$$\sigma = \mathbf{T} \mathbf{q}$$
$$\sigma_k = \mathbf{t}_k^T \mathbf{q}$$

 \square Apply "virtual load case" t_k

$$\mathbf{K}\, ilde{\mathbf{q}}_k = ilde{\mathbf{t}}_k$$

□ The first order generalized "Berke" approximation of the stress

$$\tilde{\sigma}_k = \sum_{i=1}^n \frac{d_{ik}}{x_i} \qquad \qquad d_{ik} = (\mathbf{q}_i^T \mathbf{K}_i \, \tilde{\mathbf{q}}_{ik})^0 \, x_i^0$$

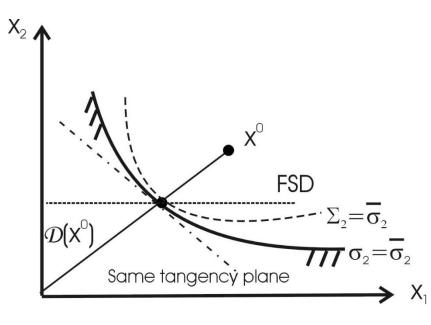
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First order explicit approximation of the stress constraints

Property:

$$\frac{\partial \sigma_k}{\partial x_i}\Big|_{\mathbf{x}^0} = -\frac{d_{ik}}{x_i^2} = \left.\frac{\partial \tilde{\sigma}_k}{\partial x_i}\right|_{\mathbf{x}^0}$$

First order approximation on the scaling line



- Approximation concept approach:
 - Linearization of the stress constraints

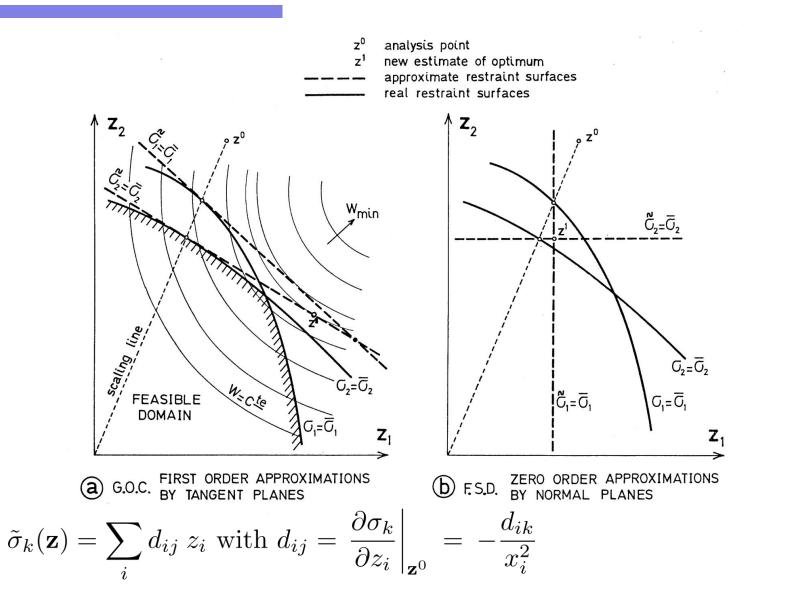
$$\sigma_k \leq \bar{\sigma}_k \quad \Rightarrow \quad \sum_i \frac{d_{ik}}{x_i} \leq \bar{\sigma}_k$$

- First order explicit approximation

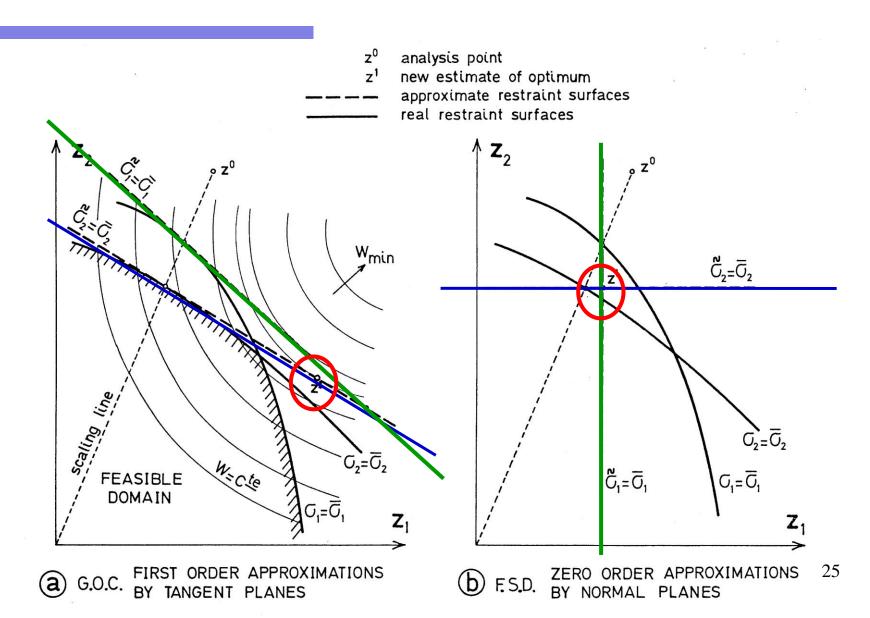
- Conventional OC: stress ratio formula
 - Fully stressed design (FSD) philosophy

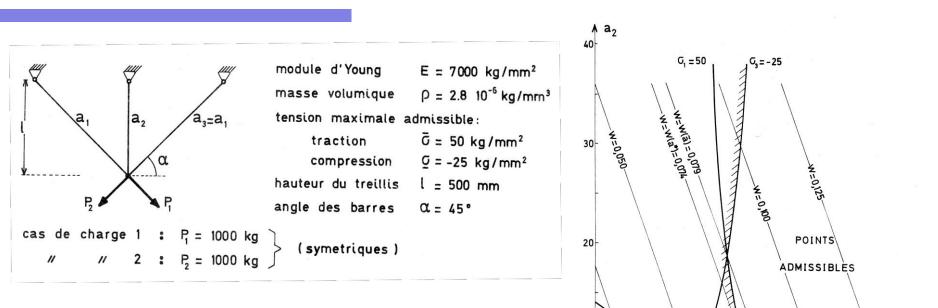
$$\sigma_k \leq \bar{\sigma}_k \implies x_k \geq \underline{\tilde{x}}_k$$
$$\underline{\tilde{x}}_k = \underline{\tilde{x}}_k^{(0)} \ \frac{\sigma_k^{(0)}}{\bar{\sigma}_k}$$

– Zero order approximation



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5:50

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_N=0,025

a * (optimum)

ã (FSD)

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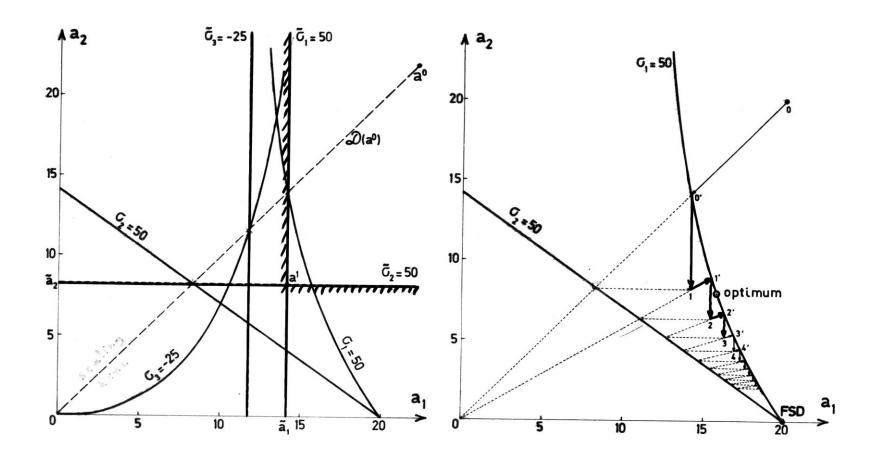
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0

Three-bar truss

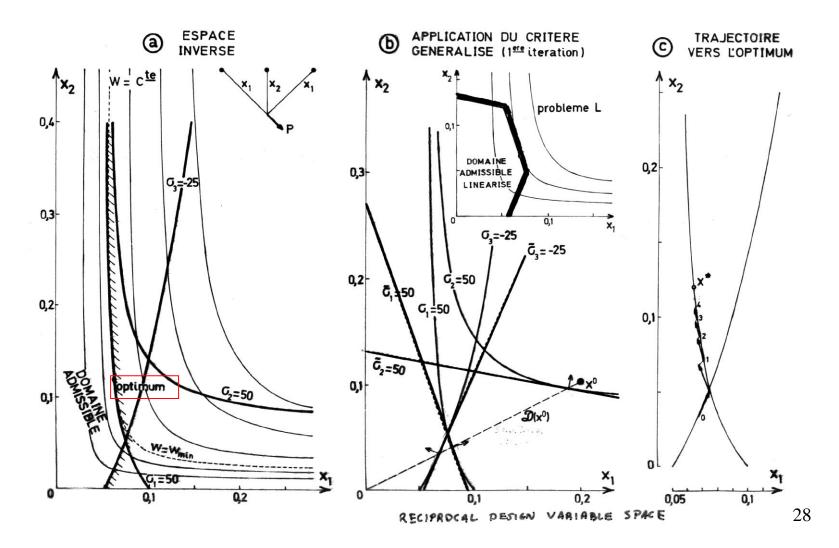
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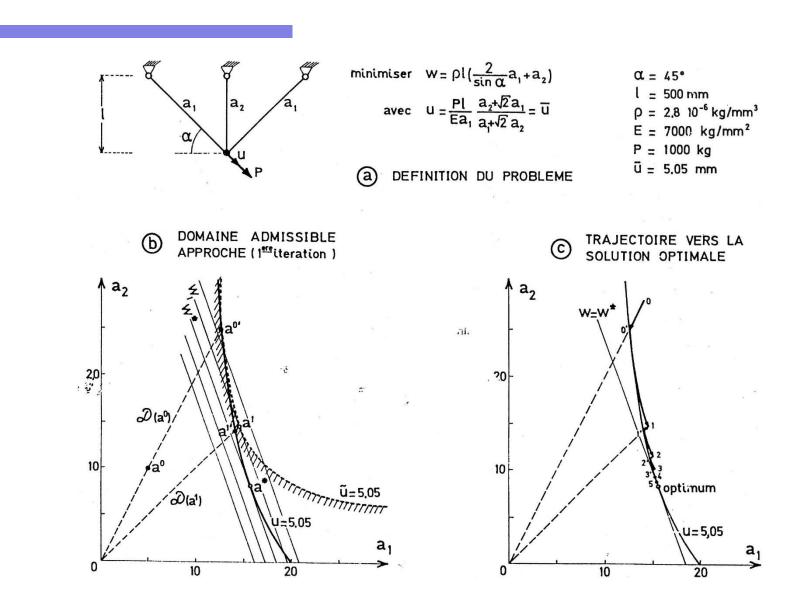
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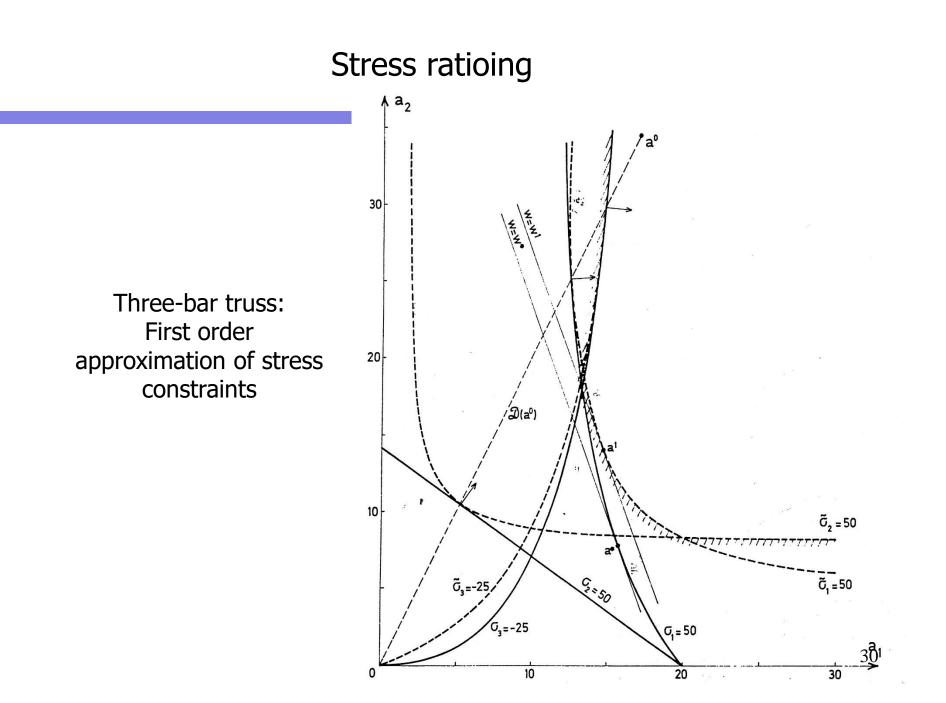
Direct design variables space

Reciprocal design variables space





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RELATIONS BETWEEN OC AND MP

- □ Generalized OC = MP linearization method
- Approximation concept approach
- Constraint gradients

Generalized OC = Linearization method

 GOC : sequence of explicit subproblems where real constraints are approximated by

$$\tilde{g}_j(x) = \sum_i \frac{c_i^0}{x_i} - \bar{u}_j \le 0$$

 $- c_{ij} = virtual energy densities$

Approximation: sequence of linearized problems with

$$\tilde{g}_{j}(z) = \left[u_{j}^{0} + \sum_{i} \frac{\partial u_{j}}{\partial z_{i}} (z_{i} - z_{i}^{0}) \right] - \bar{u}_{j} \leq 0$$
$$= \left. \sum_{i} \left. \frac{\partial u}{\partial z_{i}} \right|_{0} z_{i} - \bar{u}_{j} \leq 0 \right]$$

 $- c_{ij}$ = derivatives of the response functions with respect to the reciprocal variables

 c_{ij}^0

Generalized OC = Linearization method

- Unified approach:
 - Sequence of explicit (separable) subproblems obtained by linearizing the behavior constraints with respect to the reciprocal variables

$$\begin{array}{ll} \min & W(\boldsymbol{z}) = \sum_{i} \, \frac{w_{i}}{z_{i}} \\ \text{s.t.} & \tilde{u}_{j}(\boldsymbol{z}) = \sum_{i} \, c_{ij} \, z_{i} \, \leq \, \bar{u}_{j} \\ & \underline{z}_{i} \leq z_{i} \leq \bar{z}_{i} \end{array}$$

Later: linearizing any behavior constraints with respect to the reciprocal variables!

- Independence wrt the number of design variables!
- Solution of the explicit subproblems
 - Dual solution scheme: generalization of conventional OC techniques (GOC)
 - Primal solution scheme: mixed method: gradual transition 33
 between pure MP and OC approaches

CONCLUSION

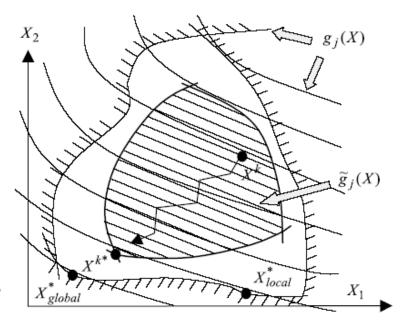
CONCLUSION

- Berke's approximation has been successful in providing high quality explicit approximations of displacement constraints
- Optimality Criteria can reduce substantially the number of function evaluation in solving costly problems in truss sizing
- They are used in building fast solution algorithms
- Berke's approximations are first order Taylor expansion of the displacement in terms of the reciprocal design variables
- How can we extend the principle to other engineering design problems?
 - Answer: Structural Approximations

SEQUENTIAL CONVEX PROGRAMMING APPROACH

Direct solution of the original optimisation problem which is generally non-linear, implicit in the design variables

$\min_{\boldsymbol{x}}$	$g_{m{0}}(m{x})$	
s.t.	$g_j(\boldsymbol{x}) \leq \bar{g}_j$	$j = 1 \dots m$



is replaced by a sequence of optimisation sub-problems

$$\min_{\boldsymbol{x}} \quad \tilde{g}_0(\boldsymbol{x})$$

s.t.
$$\tilde{g}_j(\boldsymbol{x}) \leq \bar{g}_j \quad j = 1 \dots m$$

by using approximations of the responses and using powerful mathematical programming algorithms

SEQUENTIAL CONVEX PROGRAMMING APPROACH

- Two basic concepts:
 - Structural approximations replace the implicit problem by an explicit optimisation sub-problem using convex, separable, conservative approximations; e.g. CONLIN, MMA
 - Solution of the convex sub-problems: efficient solution using dual methods algorithms or SQP method.
- Advantages of SCP:
 - Optimised design reached in a reduced number of iterations: 10 to 20 F.E. analyses
 - Efficiency, robustness, generality, and flexibility, small computation time
 - Large scale problems in terms of number of design constraints and variables
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