

# Shape Optimization

(activities 1983-2010)

Raino A. E. Mäkinen

# What is (mathematical) shape optimization ?

- In general, any optimization problem in which parameter to be optimized has some geometric interpretation (thickness, location, etc)
- In what follows, only those activities at JY related to “***PDE constrained shape optimization***” problems are considered

# Abstract setting

$$\text{Min } \mathcal{J}(\mathbf{a}) = J(\mathbf{a}, u(\mathbf{a}))$$

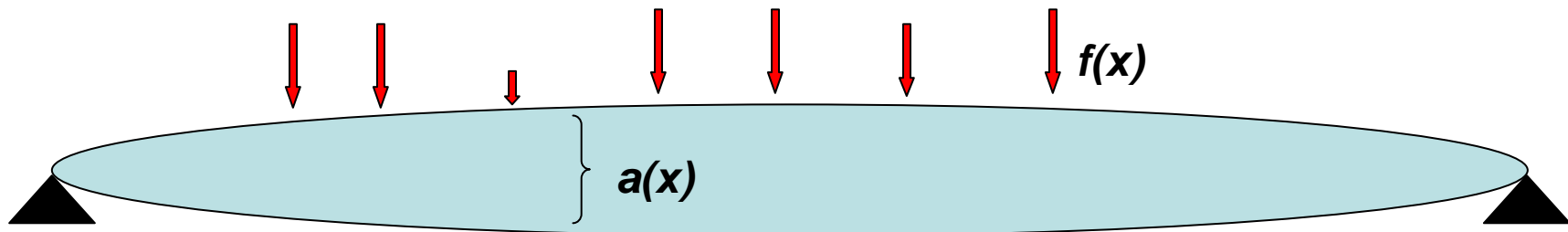
$$\mathbf{a} \in D$$

$$A(\mathbf{a})u(\mathbf{a}) = f(\mathbf{a}) \quad (\text{state problem PDE})$$

$$u(\mathbf{a}) \in S$$

- $D$  technically admissible "shapes"
- $A(\mathbf{a})$  differential operator
- $S$  admissible states of the system
- $J: D \times S \rightarrow R$  cost function

## Example: thickness optimization of a slender beam



$$\text{Min } \mathcal{J}(\mathbf{a}) = \max_x |a(x)y''(x)| \quad (\text{maximum bending stress})$$

$$\text{s.e. } \int a(x)dx = V \quad \text{ja } a(x) \geq a_0 > 0$$

$$([a(x)]^3 y''(x))'' = f(x) \quad 0 < x < L$$

$$y(0) = y''(0) = y(L) = y''(L) = 0$$

Control parameter is the function  $\mathbf{a}: [0, L] \rightarrow \mathbf{R}$

# Multiphysics and/or multicriteria shape optimization

$$\text{Min}_{a \in D} \{ J_1(a, u(a)), J_2(a, w(a)) \}$$

$$a \in D$$

$$A_1(a)u(a) = f_1(a)$$

$$A_2(a)w(a) = f_2(a)$$

$$u(a) \in S_1, w(a) \in S_2$$



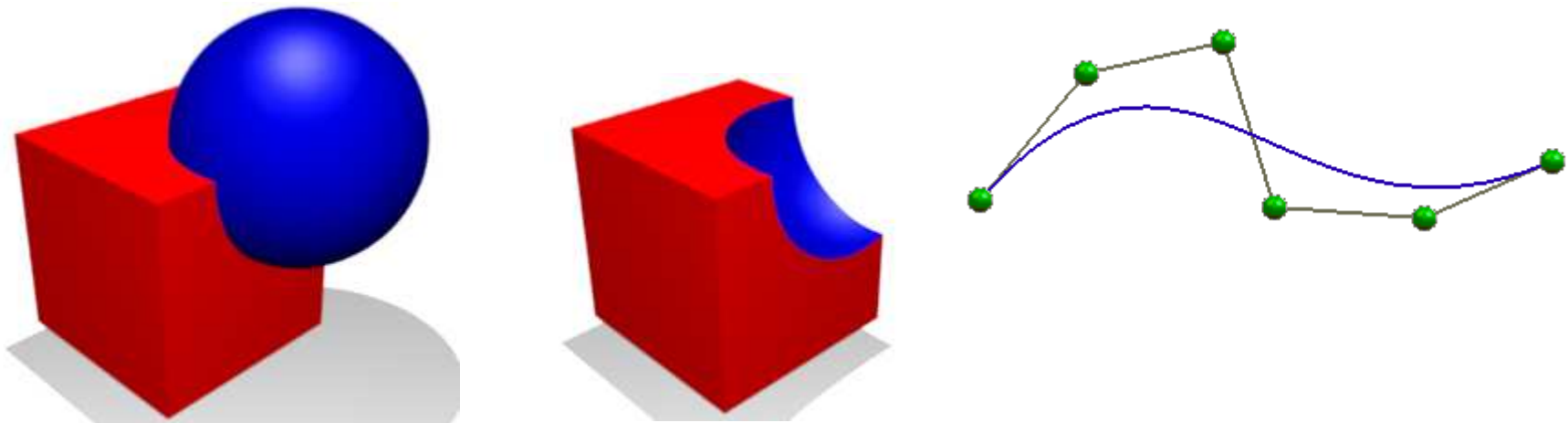
Examples:

- Aeroelasticity (fluid + structure)
- Stealth (fluid + EM / acoustics)
- Mobile technology (structure + EM + heat + acoustics)



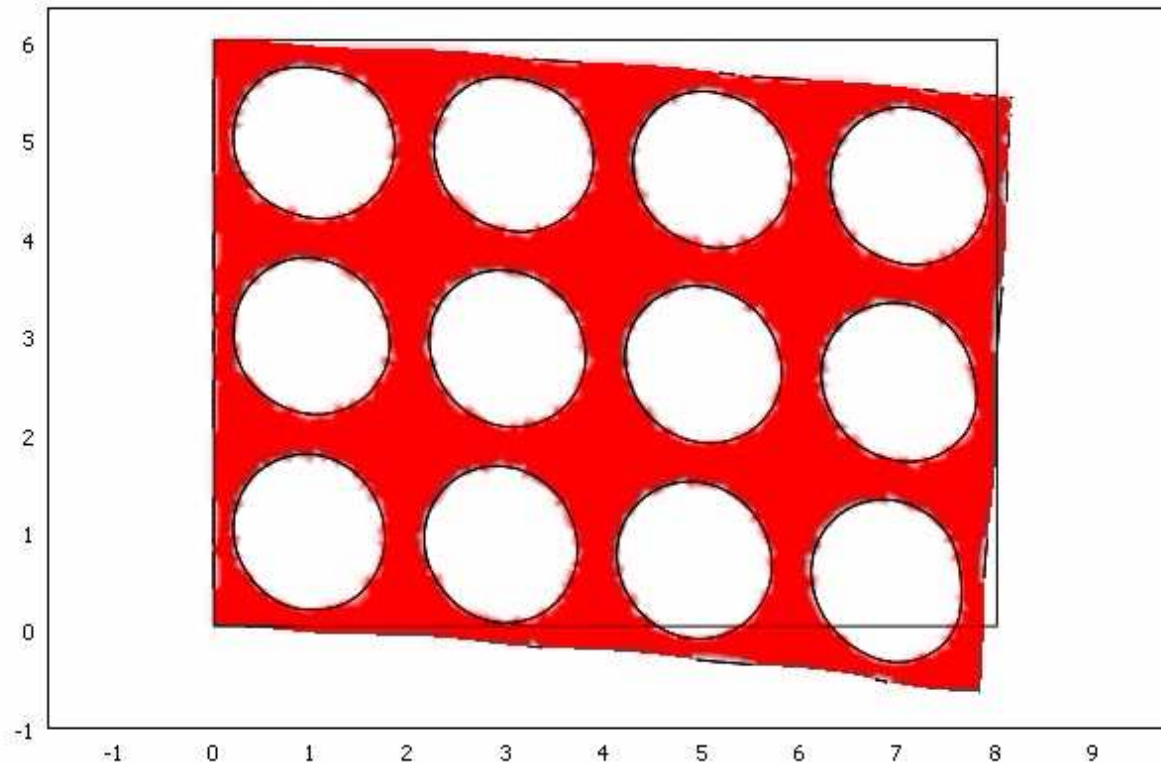
# Shape parameterization

- CAD-parameterization
  - Eg. CSG, length, thickness, radius, Spline curve control points,...



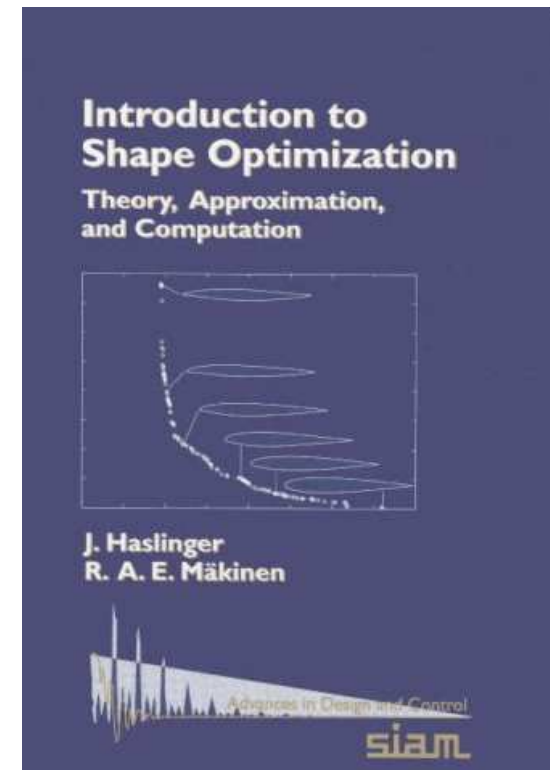
- “CAD-free” parameterization
  - “Move” computational grid
- “Level-set function” parameterization:  
 $\varphi: \mathbb{R}^d \rightarrow \mathbb{R}$ ,  $\Omega := \{ x: \varphi(x) > 0 \}$ ,  $\varphi(x) = \sum a_i \varphi_i(x)$

# Example: Topology optimization of a cantilever using “level set” method



# Past and recent activities in Shape Optimization

- **First journal paper:** J. Haslinger, P. Neittaanmäki: *Penalty method in design optimization of systems governed by mixed Dirichlet - Signorini boundary value problem*, Ann. Fac. Sci. Toulouse, 199-216, **1983**.
- **First PhD Thesis:** by T. Tiihonen, **1987**
  - Others: Mäkinen, Salmenjoki, Jari Toivanen, (Kari Kärkkäinen)
- **Monographs:**
  - Haslinger & Neittaanmäki, 1988
  - Haslinger & Neittaanmäki, 1996
  - Neittaanmäki, Rudnicki & Savini, 1996
  - Haslinger & Mäkinen, 2003
  - Banichuk & Neittaanmäki, 2009
- **Papers:**
  - Over 25 journal articles in ISI Web of Science
- **Collaboration with industry:**
  - Technology transfer





# Recent journal papers by RM et al.

- Toivanen, Mäkinen, Rahola, Järvenpää & Ylä-Oijala (2010). **Gradient-based shape optimisation of ultra-wideband antennas parameterised using splines**. IET Microwaves, Antennas & Propagation.
- Toivanen & Mäkinen (2010). **Implementation of sparse forward mode automatic differentiation with application to electromagnetic shape optimization**. Optimization Methods & Software
- Toivanen, Mäkinen, Järvenpää, Ylä-Oijala & Rahola (2009). **Electromagnetic Sensitivity Analysis and Shape Optimization Using Method of Moments and Automatic Differentiation**. IEEE Antennas and Propagation.
- Toivanen, Haslinger & Mäkinen (2008). **Shape optimization of systems governed by Bernoulli free boundary problems**. Computer Methods in Applied Mechanics and Engineering.
- Stebel, Mäkinen & Toivanen (2007). **Optimal shape design in a fibre orientation model**. Applications of Mathematics.

# Adjoint equation method

$$\begin{aligned} \text{Min } \mathcal{J}(\mathbf{a}) &= J(\mathbf{a}, u(\mathbf{a})) \\ A(\mathbf{a})u(\mathbf{a}) &= f(\mathbf{a}) \end{aligned}$$

How to compute  $\partial \mathcal{J}(\mathbf{a}) / \partial \mathbf{a}_k$  at  $\mathbf{a}$  ?

1) Solve state problem:  $K(\mathbf{a})u = f(\mathbf{a})$

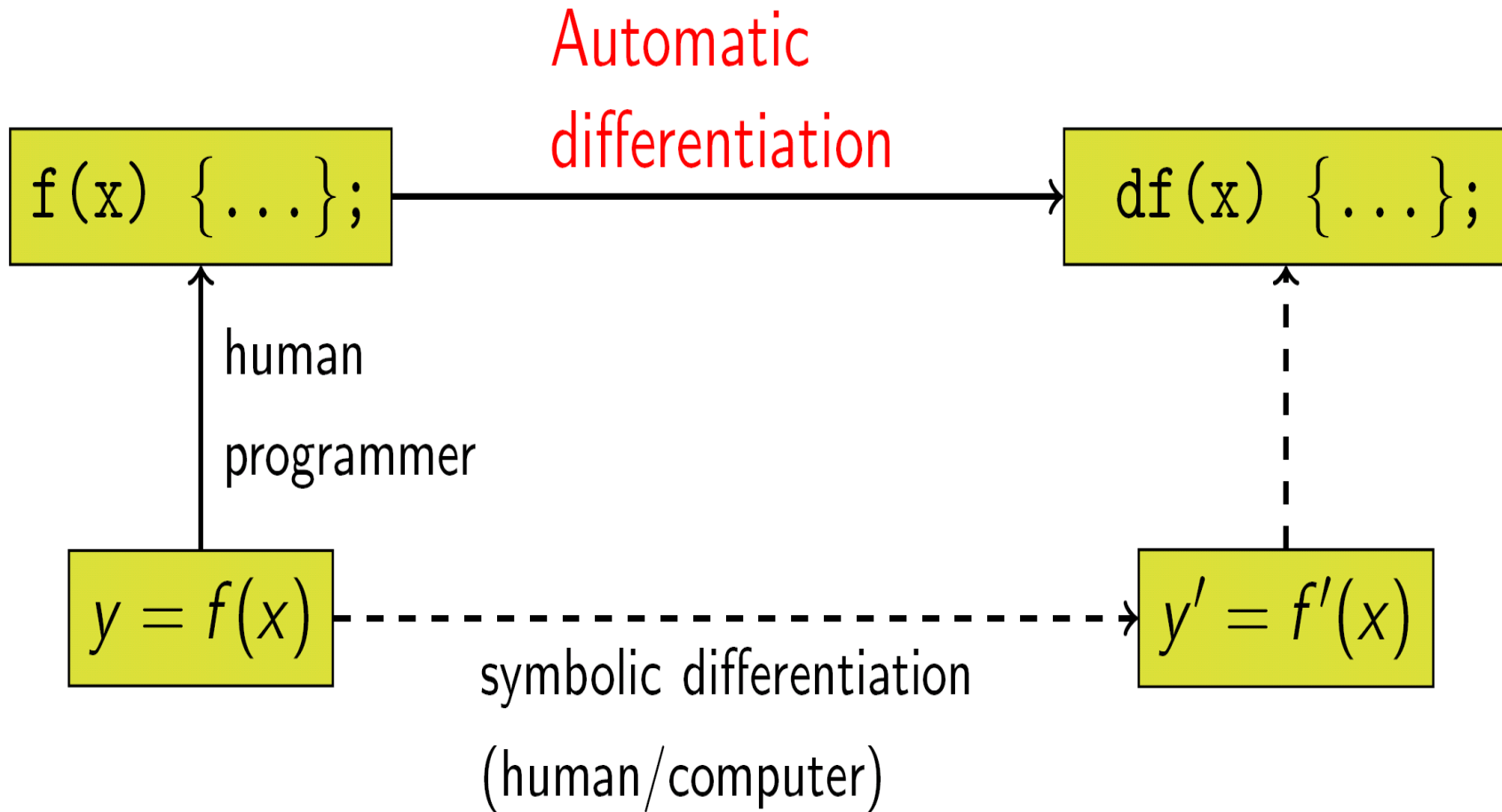
2) Solve adjoint equation:

$$[A(\mathbf{a})]^T \mathbf{p} = \partial J(\mathbf{a}, u) / \partial u$$

3) Compute partial derivative:

$$\partial \mathcal{J}(\mathbf{a}) / \partial \mathbf{a}_k = \mathbf{p}^T [ \partial f(\mathbf{a}) / \partial \mathbf{a} - (\partial K(\mathbf{a}) / \partial \mathbf{a})u ]$$

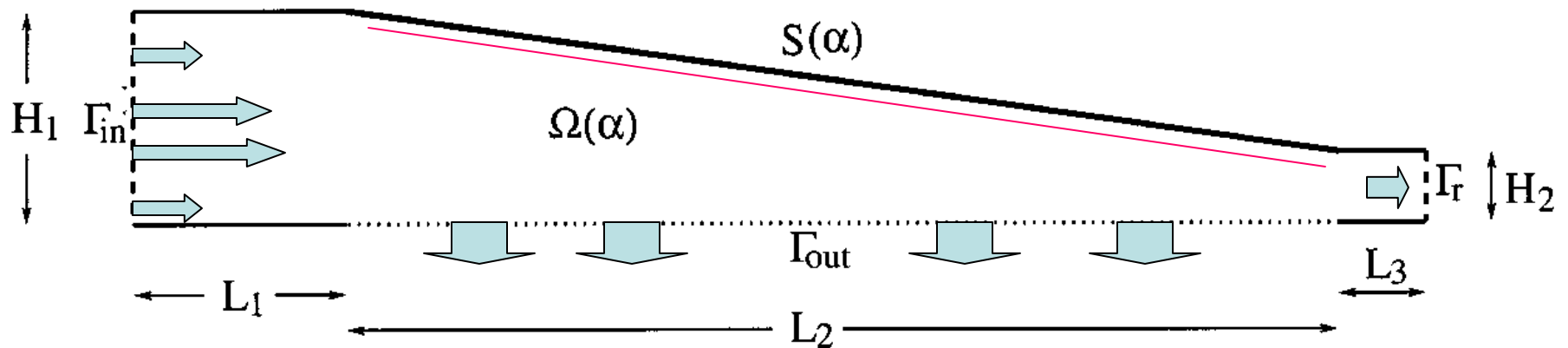
# Automatic vs. symbolic differentiation



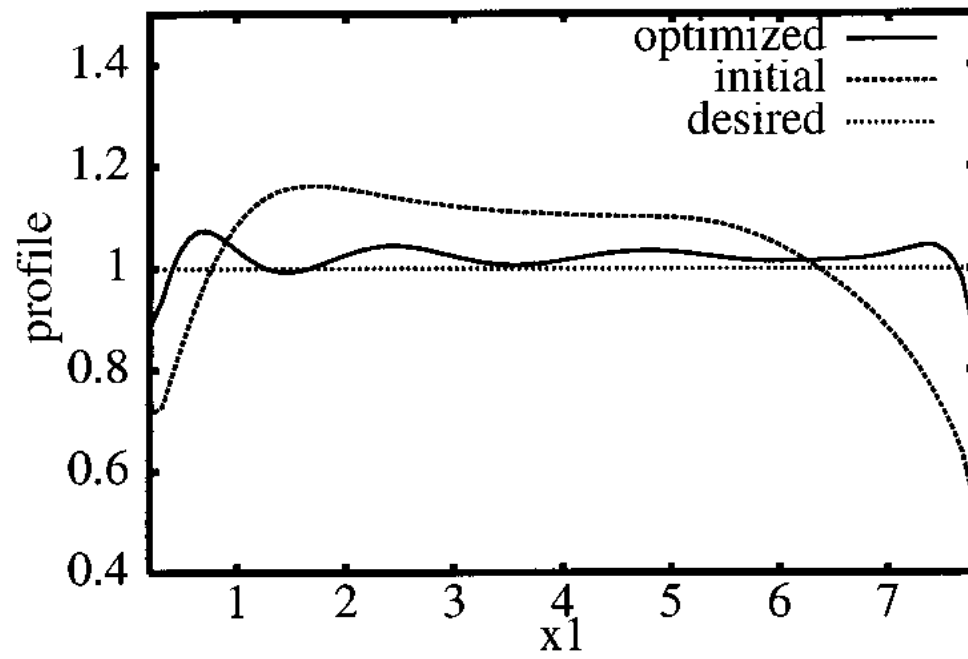
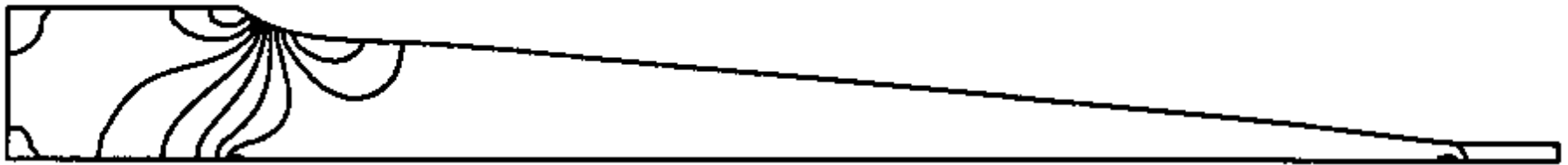
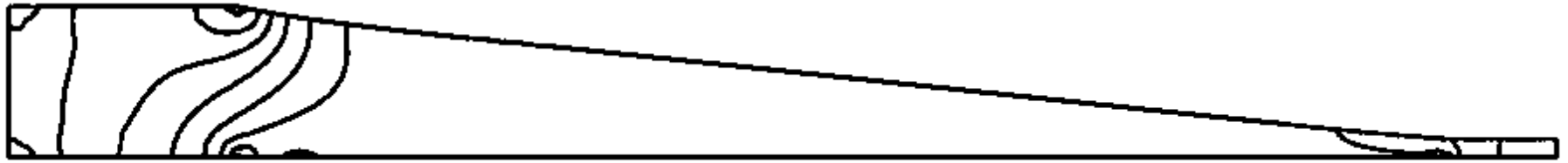
# AD in shape optimization

- In principle, one could apply AD to whole (FEM) simulation program
- In practice, often only certain sensitivities (e.g.  $\partial \mathbf{K}(\mathbf{a})/\partial \mathbf{a}$  and  $\partial \mathbf{f}(\mathbf{a})/\partial \mathbf{a}$ ) are computed using AD.
- Derivatives computed using AD very useful for the simulation model alone (Jacobian for Newton method in case of nonlinear state problem)!!

# Example: Shape optimization of a dividing tube

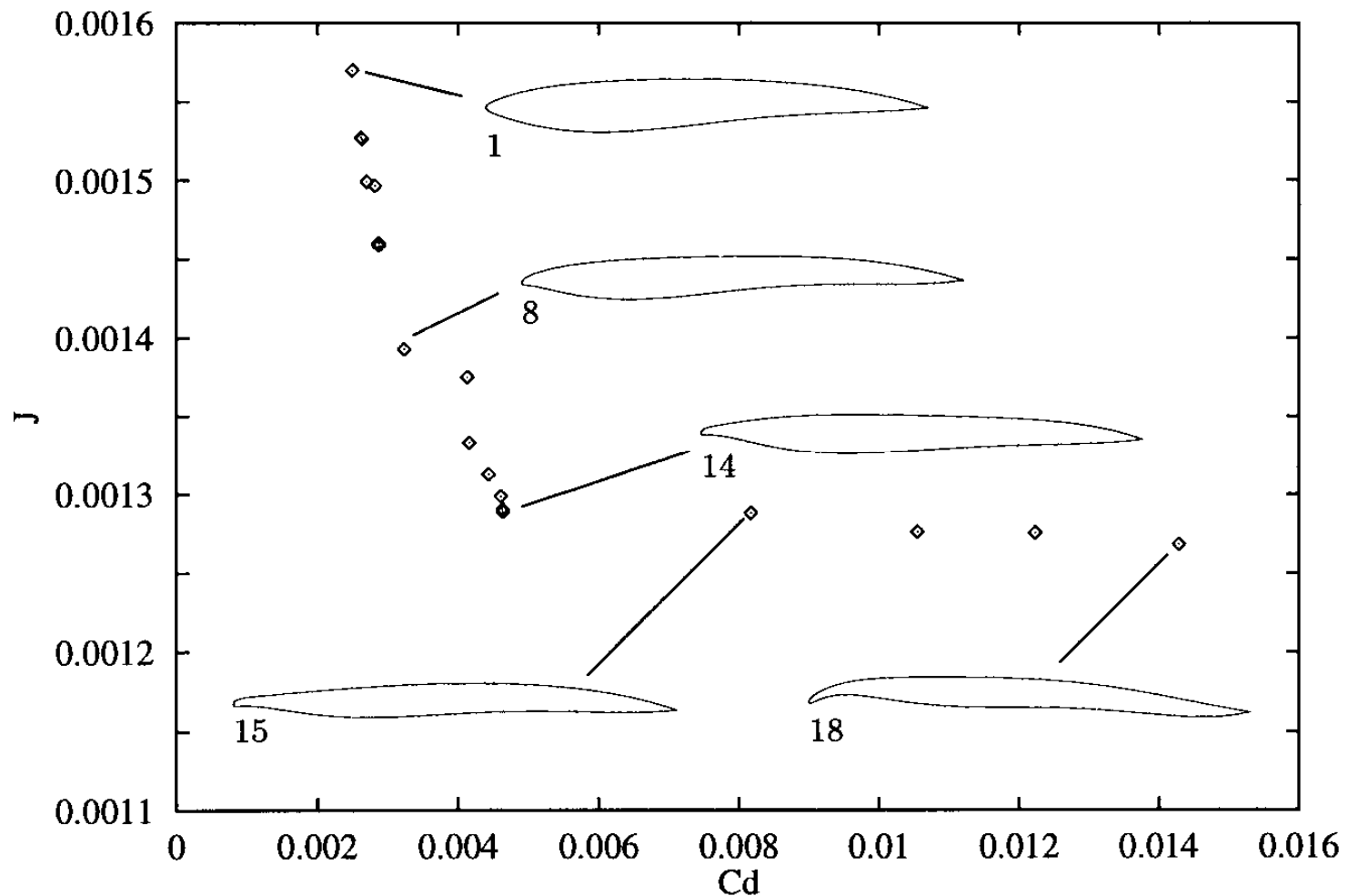


- Navier-Stokes equation, algebraic turbulence model
- FEM, analytic sensitivity analysis for discretized state problem using hybrid AD/hand coded derivatives
- Shape optimization using gradient method (SQP) needs ca. 20 function evaluations (= FEM simulations).



# Example: Multiphysics multicriteria shape optimization of an airfoil using a Genetic Algorithm

Optimization of Radar Cross Section (RCS) and drag coefficient ( $C_D$ ) subject to constraint for lift coefficient ( $C_L$ ). CFD-model: Euler, CEM-model: Helmholtz. Optimization takes ca. 6000 function evaluations ☹



# Example: Shape optimization of patch dipole antenna

Surface integral equation formulation of 3D time-harmonic Maxwell's equations:

$$\left( -\frac{1}{i\omega\epsilon_0} \nabla \mathbf{S}(\nabla_S \cdot \mathbf{J})(\mathbf{r}) + i\omega\mu_0 \mathbf{S}(\mathbf{J})(\mathbf{r}) \right)_{\text{tan}} = -\mathbf{E}_{\text{tan}}^p(\mathbf{r})$$

$$\mathbf{S}(\mathbf{F})(\mathbf{r}) = \int_S G_0(\mathbf{r}, \mathbf{r}') \mathbf{F}(\mathbf{r}') dS'$$

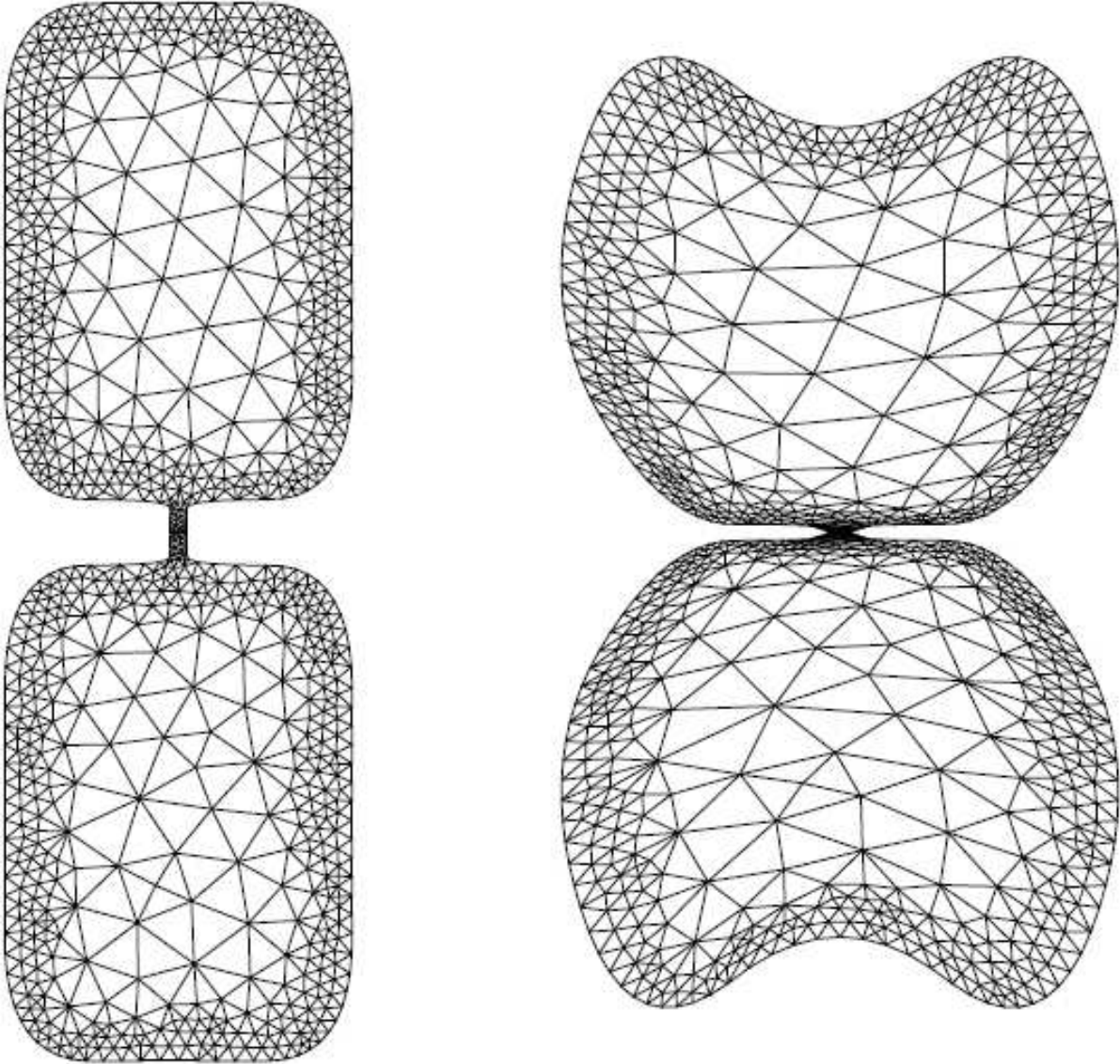
Method of Moments discretization leads to a large **dense** linear system:

$$\mathbf{A}\mathbf{x} = \mathbf{b}$$

We wish to design an antenna that performs well over wide frequency range (multipoint design) by minimizing "S-parameter".



Initial versus optimal shape (and computational mesh)



# Performance over frequency range

