

TRUSS SIZING AND SHAPE OPTIMIZATION: A SECOND-ORDER APPROACH

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OPTIMIZATION ALGORITHMS

- Genetic algorithms
 - ◆ Derivative free
 - ◆ Robust in global optimization
 - ◆ Can be easily parallelized
 - ◆ Inefficient when the number of variables is high

OPTIMIZATION ALGORITHMS (cont.)

- First order methods
 - ◆ Structural analysis / Sensitivity analysis / Redesign
 - ◆ First order sensitivity analysis
 - ◆ Adequate for a moderate number of design variables

OPTIMIZATION ALGORITHMS (cont.)

- Second order method presented here
 - ◆ Integrated formulation
 - ◆ First and second derivatives are symbolically determined
 - ◆ Adequate for problems with a large number of design variables
 - ◆ Penalized by the presence of a large number of behavior variables

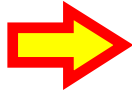
NONLINEAR PROGRAMMING

Minimize $f(\underline{x})$

subject to

$$g(\underline{x}) \leq \underline{0} \quad \rightarrow \quad g_i(\underline{x}) + s_i^2 = 0$$

$$h(\underline{x}) = \underline{0}$$

- Variables / functions  real and continuous
- All the functions are generalized polynomials, such as:

$$f(\underline{x}) = 5.9x_1^2x_4^{-3} - 3.1x_2 + 2.7x_1^{-1}x_3x_5^2 - 1.8$$

GENERALIZED POLYNOMIALS

$$f(\underline{x}) = 5.9x_1^2x_4^{-3} - 3.1x_2 + 2.7x_1^{-1}x_3x_5^2 - 1.8$$

- A symbolic manipulation is performed
- Expression parsing and evaluation is simplified
- Exact first and second derivatives can be easily calculated
- All these operations can be efficiently performed

INPUT FILE

```
### Main title of the nonlinear program
      Symmetric truss with two load cases (kN,cm)
Min.
      +565.685 * t5 ^ 2 + 100 * t8 ^2 ; # truss volume (cm3)

s.t.i.c.
      Min. area 4:   - t4 ^ 2 + 0.15 < 0 ;

s.t.e.c.
      Equil 16:   + 141.421 * t5 ^ 2 * disp16 - 100 = 0 ;

END_OF_FILE
```

- All the software is coded in ANSI C

LAGRANGIAN

$$L(\underset{\sim}{X}) = f(\underset{\sim}{x}) + \sum_{k=1}^m \lambda_k^g \left[g_k(\underset{\sim}{x}) + s_k^2 \right] + \sum_{k=1}^p \lambda_k^h h_k(\underset{\sim}{x})$$

VARIABLES

$$\underset{\sim}{X} = \left(\underset{\sim}{s}, \underset{\sim}{\lambda}^g, \underset{\sim}{x}, \underset{\sim}{\lambda}^h \right)$$

SOLUTION

- Stationary point of the Lagrangian

SYSTEM OF NONLINEAR EQUATIONS

$$\nabla L(\tilde{X}) = \tilde{0} \Rightarrow \begin{cases} 2s_i\lambda_i^g = 0 & (i = 1, \dots, m) \\ g_i + s_i^2 = 0 & (i = 1, \dots, m) \\ \frac{\partial f}{\partial x_i} + \sum_{k=1}^m \lambda_k^g \frac{\partial g_k}{\partial x_i} + \sum_{k=1}^p \lambda_k^h \frac{\partial h_k}{\partial x_i} = 0 & (i = 1, \dots, n) \\ h_i = 0 & (i = 1, \dots, p) \end{cases}$$

- The solution of the system is a KKT solution when

$$\lambda_{\tilde{}}^g \geq 0$$

LAGRANGE-NEWTON METHOD

- The system of nonlinear equations

$$\nabla L(\tilde{X}) = \tilde{0}$$

is solved by the Newton method

- In each iteration the following system of linear equations has to be solved

$$H\left(\tilde{X}^{q-1}\right) \Delta \tilde{X}^q + \nabla L\left(\tilde{X}^{q-1}\right) = \tilde{0}$$

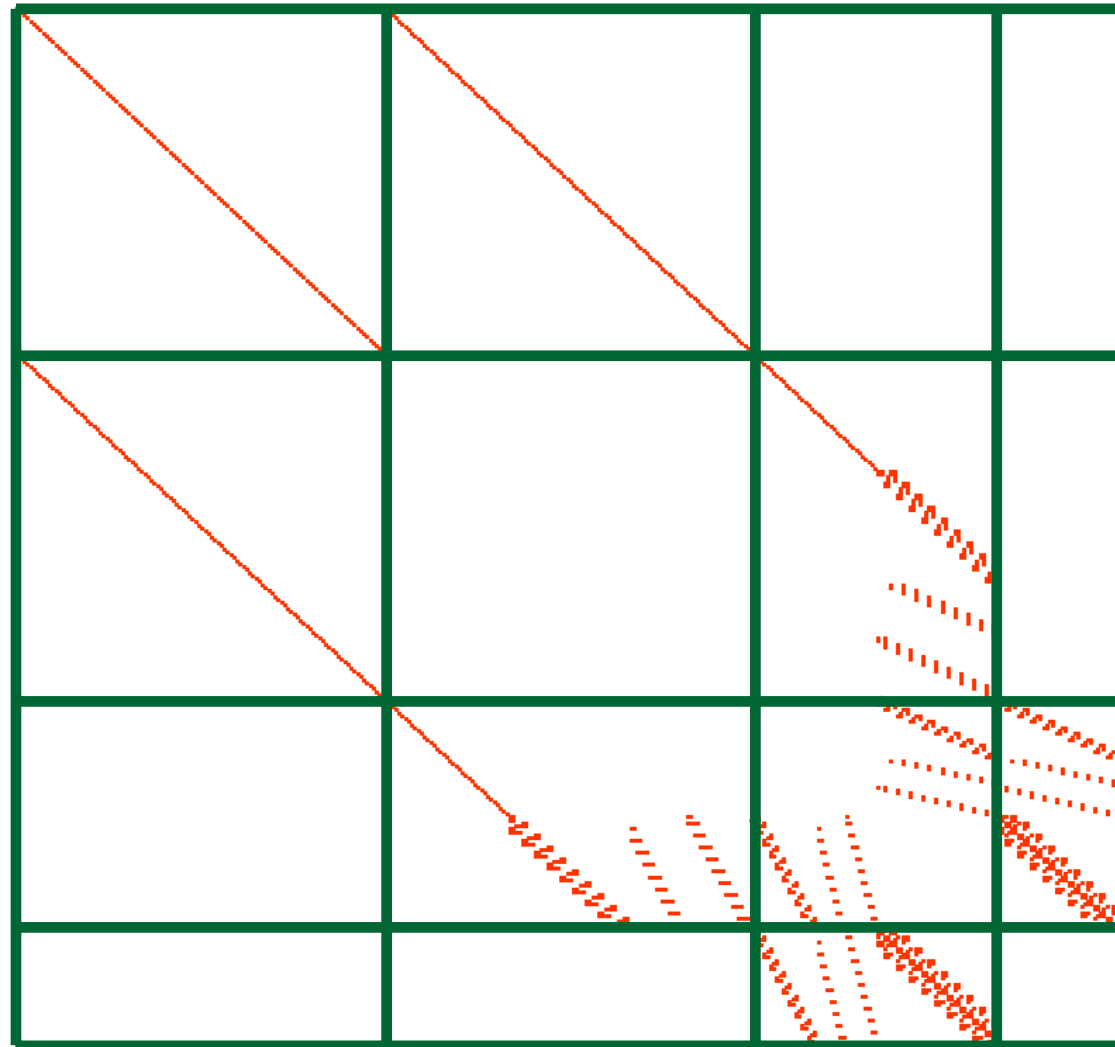
HESSIAN MATRIX

$$\tilde{H} = \begin{array}{c} \begin{array}{c} (m) \\ (m) \\ (n) \\ (p) \end{array} \begin{array}{c} \begin{array}{c} (m) \\ (m) \\ (n) \\ (p) \end{array} \end{array} \begin{array}{|c|c|c|c|} \hline \text{Diag}(2\lambda_i^g) & \text{Diag}(2s_i) & 0 \\ \hline & 0 & \frac{\partial g_i}{\partial x_j} \\ \hline & & \bullet & \frac{\partial h_j}{\partial x_i} \\ \hline \text{SYMMETRIC} & & & 0 \\ \hline \end{array} \end{array}$$

$$\bullet \frac{\partial^2 f}{\partial x_i \partial x_j} + \sum_{k=1}^m \lambda_k^g \frac{\partial^2 g_k}{\partial x_i \partial x_j} + \sum_{k=1}^p \lambda_k^h \frac{\partial^2 h_k}{\partial x_i \partial x_j}$$

HESSIAN MATRIX SPARSITY PATTERN

$H =$
 \sim



SYSTEM OF LINEAR EQUATIONS

- Gaussian elimination
 - ◆ adapted to the sparsity pattern of the Hessian matrix
- Conjugate gradients
 - ◆ diagonal preconditioning
 - ◆ adapted to an indefinite Hessian matrix

LINE SEARCH

$$\tilde{X}^q = \tilde{X}^{q-1} + \alpha \Delta \tilde{X}^q$$

- When the value of α minimizes the error in $\Delta \tilde{X}^q$ direction
 - ♦ the value of α is often close to one
 - ♦ faster convergence
 - ♦ process may fail
- When the value of α is made considerably smaller (e.g. $\alpha = 0.1$)
 - ♦ stable convergence
 - ♦ more iterations - slower

***NEWTOP* COMPUTER CODE**

- All the variables are scaled
- Constraints are normalized
- Elementary equality constraints are substituted:

$$x_i = c x_j \quad \text{or} \quad x_i = c$$

- The NLP is simplified
- Problems with a large number of variables can be solved
(e.g., 4 000 design variables and 20 000 constraints)

TRUSS OPTIMIZATION

- Cost minimization (often similar to volume minimization)

- Sizing \Rightarrow cross-sectional areas may change

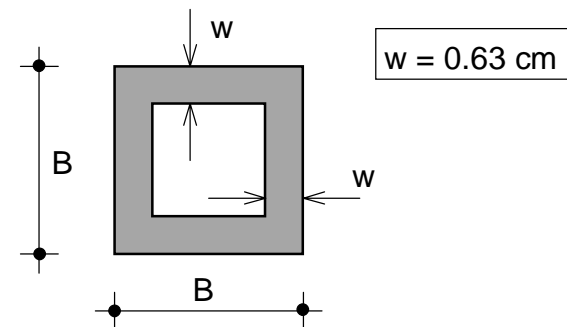
- Shape optimization \Rightarrow nodal coordinates may change



Simultaneously

VARIABLES

- Integrated formulation
- Design variables and behavior variables simultaneously present in the nonlinear program
 - ◆ Cross-section dimensions (e.g., width, diameter, area)
 - ◆ Some nodal coordinates
 - ◆ Nodal displacements



SUBSTITUTED VARIABLES

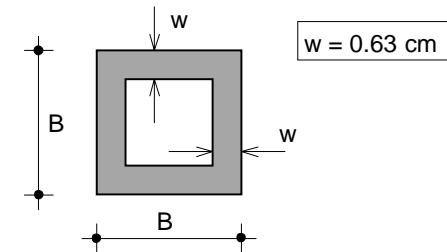
- In most cases the area (**A**) and the moment of inertia (**I**) depend on a single parameter (**B**)

$$A = C_0^A + C_1^A B + C_2^A B^2$$

$$I = C_0^I + C_1^I B + C_2^I B^2 + C_3^I B^3 + C_4^I B^4$$

◆ Coefficients C_i^A and C_j^I are fixed

◆ Variables **A** and **I** can be substituted in all the functions that define the mathematical program

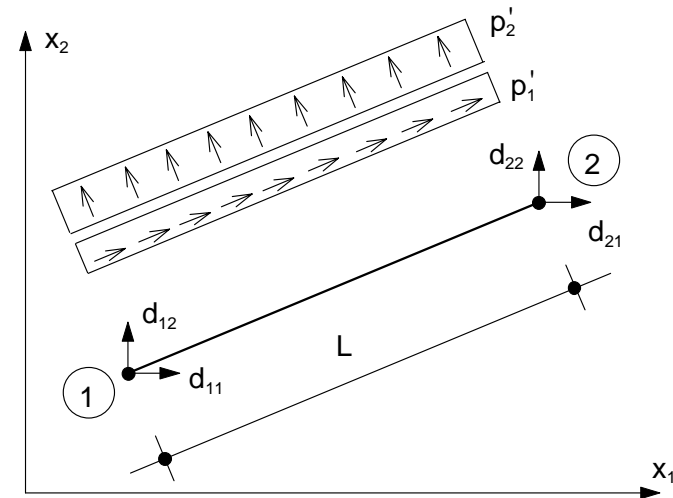


ADDITIONAL VARIABLES

$$k_{ij} = \dots + EAL^{-1} + \dots$$

$$L = \sqrt{(x_{21} - x_{11})^2 + (x_{22} - x_{12})^2}$$

◆ Additional variables $\Rightarrow L_i$



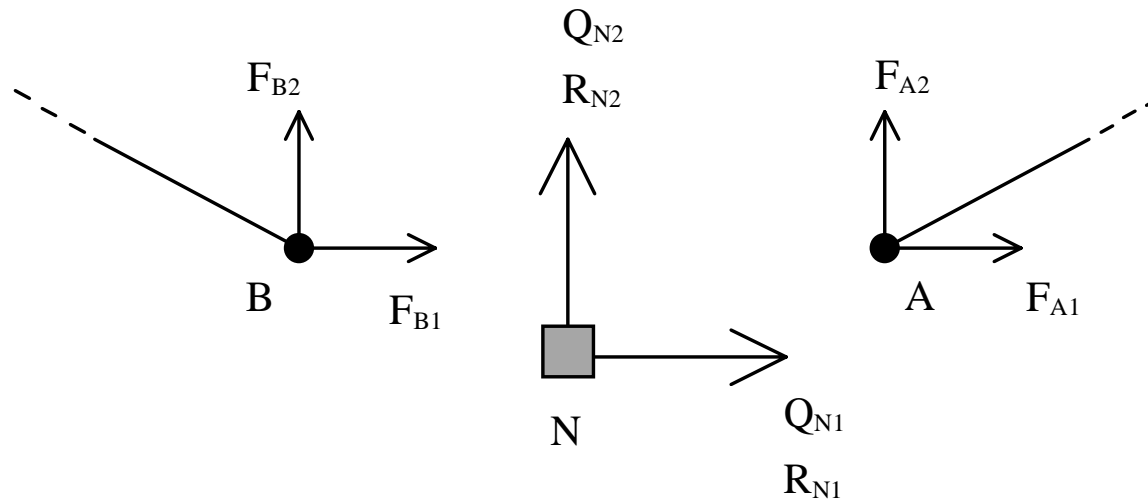
$$-L^2 + x_{11}^2 + x_{12}^2 + x_{21}^2 + x_{22}^2 - 2x_{11}x_{21} - 2x_{12}x_{22} = 0$$

◆ Additional equality constraints $\Rightarrow L_i$ definition

EQUILIBRIUM EQUATIONS

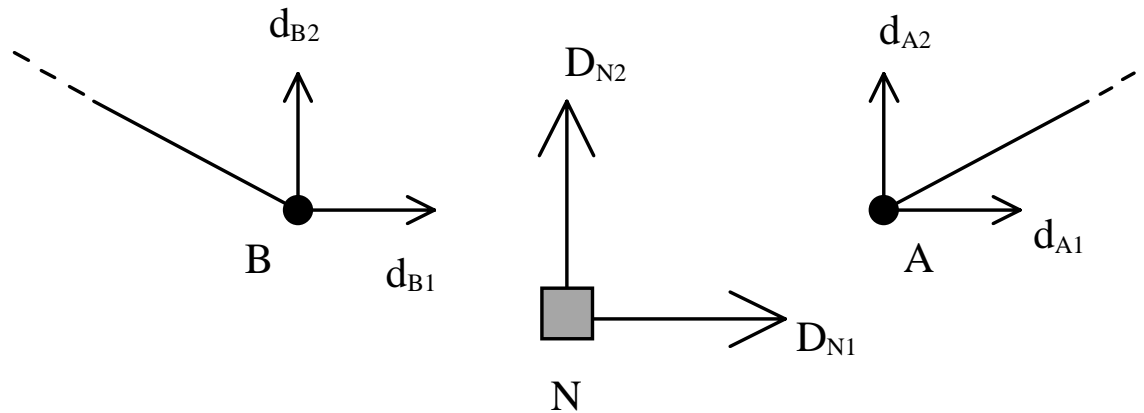
- Equality constraints:

$$\tilde{F}_A + \dots + \tilde{F}_B + \dots = \tilde{Q} + \tilde{R}$$



- Reactions are only present in constrained dof 's

COMPATIBILITY EQUATIONS



$$\begin{aligned} \tilde{d}_A &= \tilde{D}_N \\ \tilde{d}_B &= \tilde{D}_N \end{aligned}$$

- Variables \mathbf{d} are substituted
- D_{Ni} is fixed in constrained dof 's

NON LINEAR PROGRAM

- Objective function: cost $\Rightarrow f(\underline{x}) = \sum_{i=1}^{NB} c_i A_i L_i$
- Equality constraints:
 - ♦ for each bar with variable length:
 - \rightarrow one equation defining L
 - ♦ for each non-prescribed degree of freedom:
 - \rightarrow one equilibrium equation

- Inequality constraints:

- ♦ minimum width $\Rightarrow B \geq B_{\min}$

- ♦ allowable stress (tension and compression)

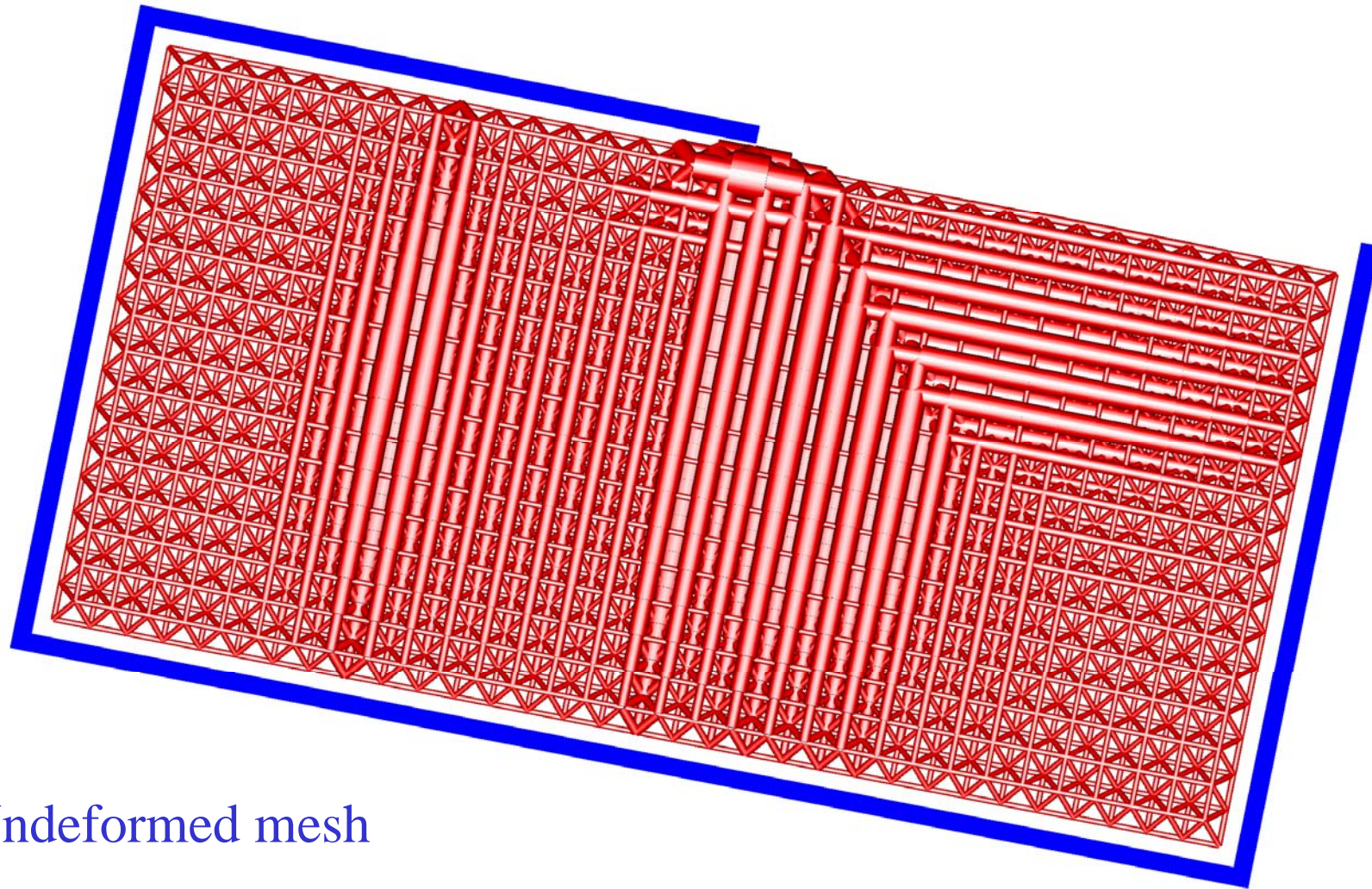
- ♦ local Euler buckling

- ♦ side constraints in nodal coordinates $\Rightarrow x_{\min} \leq x_i \leq x_{\max}$

LARGE SCALE OPTIMIZATION PROBLEM

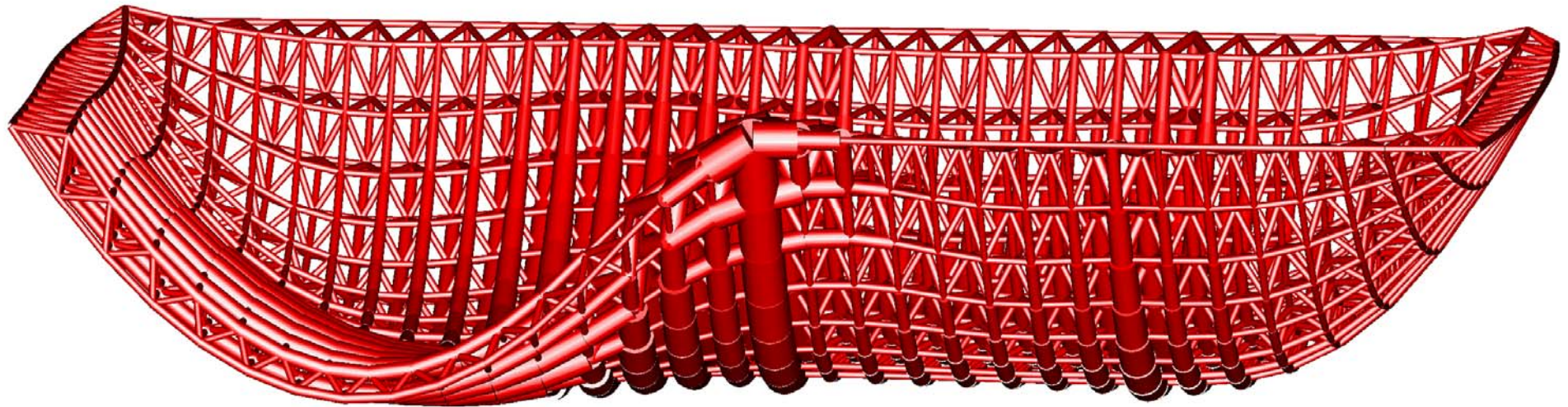
- 3D truss sizing
- Number of bars = 4 096
- Number of degrees of freedom = 3 135
- Number of decision variables = 7 231
- Number of inequality constraints = 19 038
- No variable linking
- No active set strategy

BUILDING ROOF - OPTIMAL SOLUTION



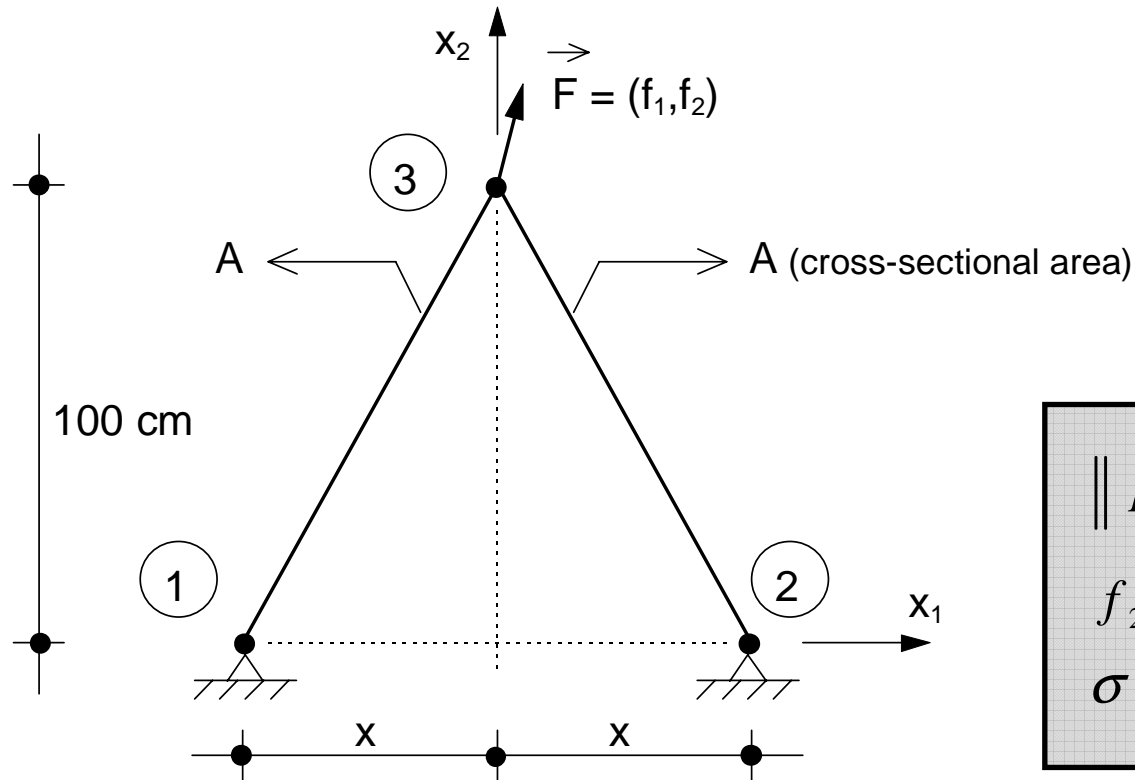
Undeformed mesh

BUILDING ROOF - OPTIMAL SOLUTION



Deformed mesh

SHAPE OPTIMIZATION TEST PROBLEM



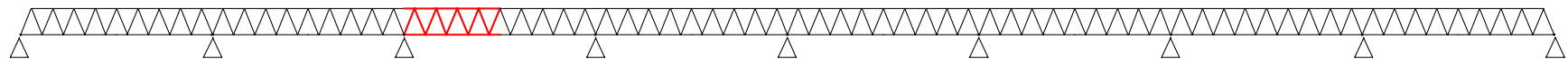
$$\begin{aligned} \|\vec{F}\| &= 200 \text{ kN} \\ f_2 &= 8 f_1 \\ \sigma_{\max} &= 100 \text{ kN/cm}^2 \end{aligned}$$

• Variables: A , x

• Svanberg's solution confirmed

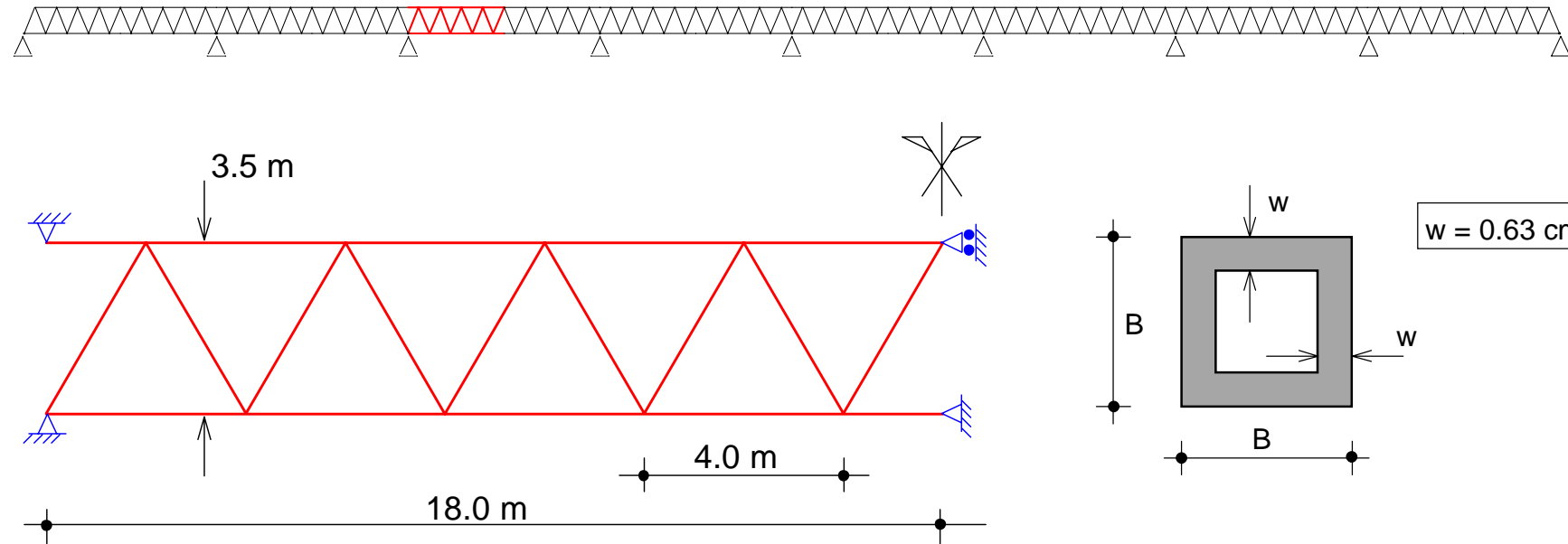
SHAPE OPTIMIZATION PROBLEM

- Minimize the cost of a steel bridge
- Member sizing and shape optimization
- Linear elastic structural behavior
- Fixed nodes (normal direction)
- Local Euler buckling
- Portuguese structural codes



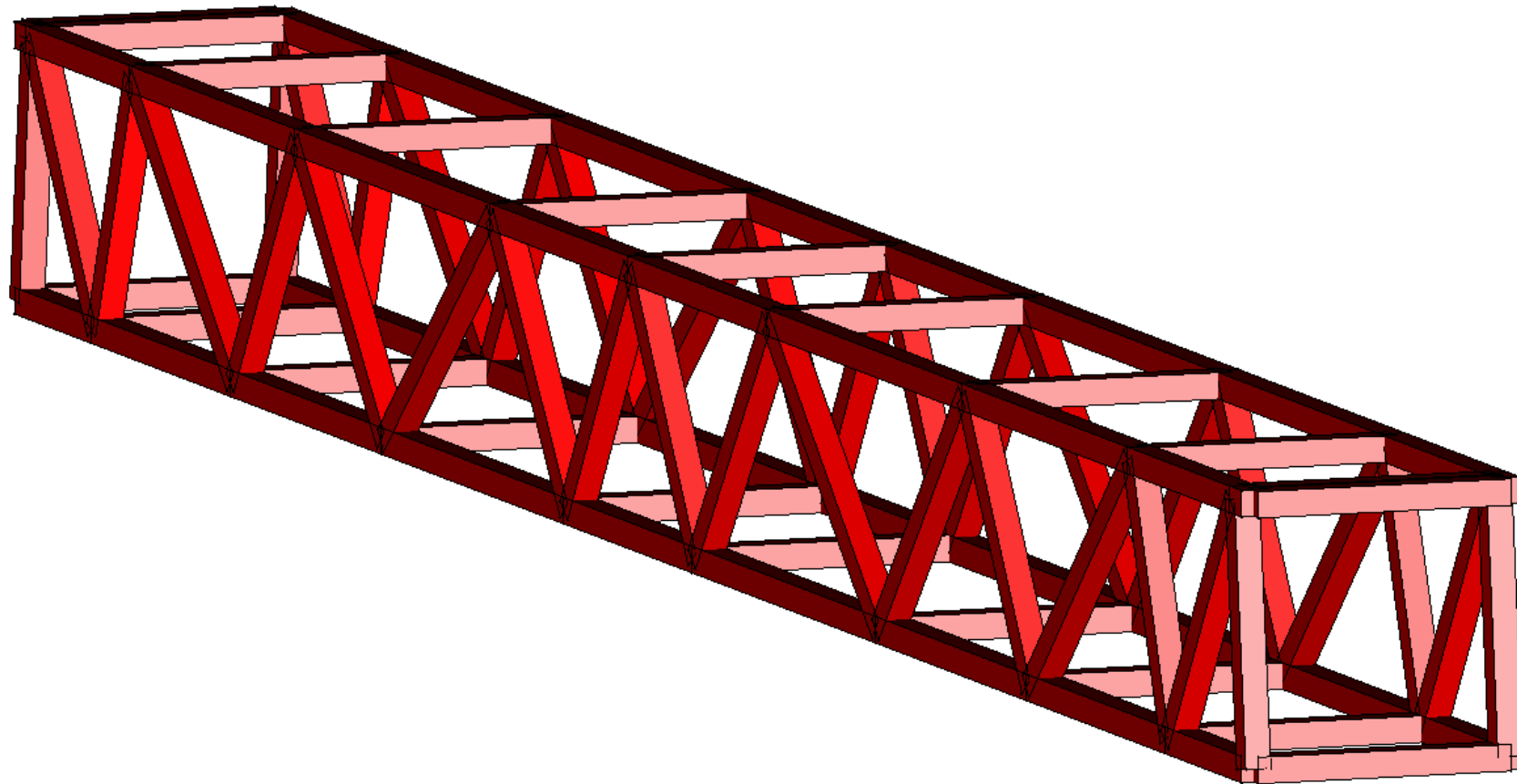
STEEL BRIDGE

Vertical distributed load

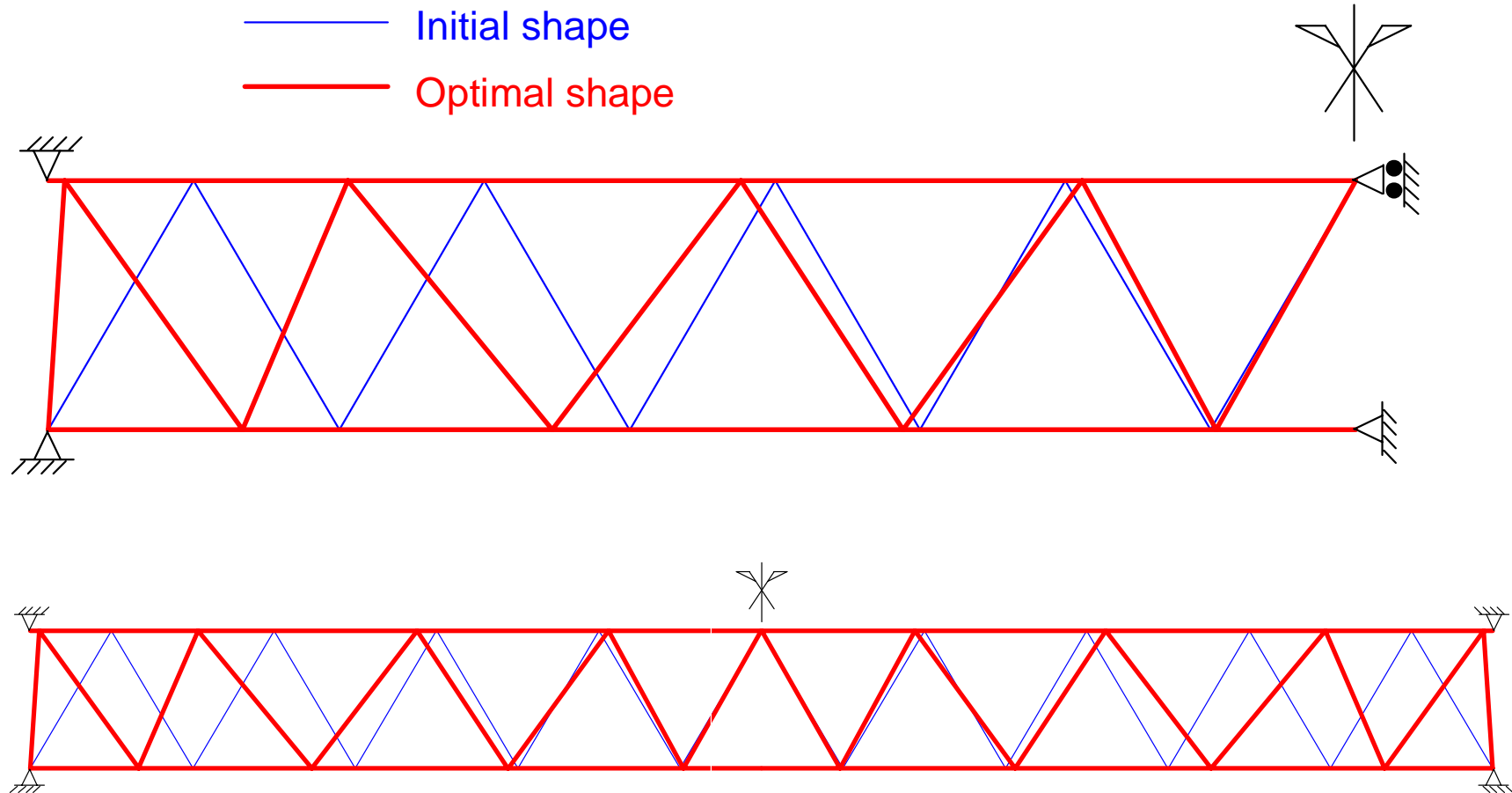


- Group I - horizontal bars
- Group II - diagonal bars

STEEL BRIDGE



OPTIMAL SHAPE

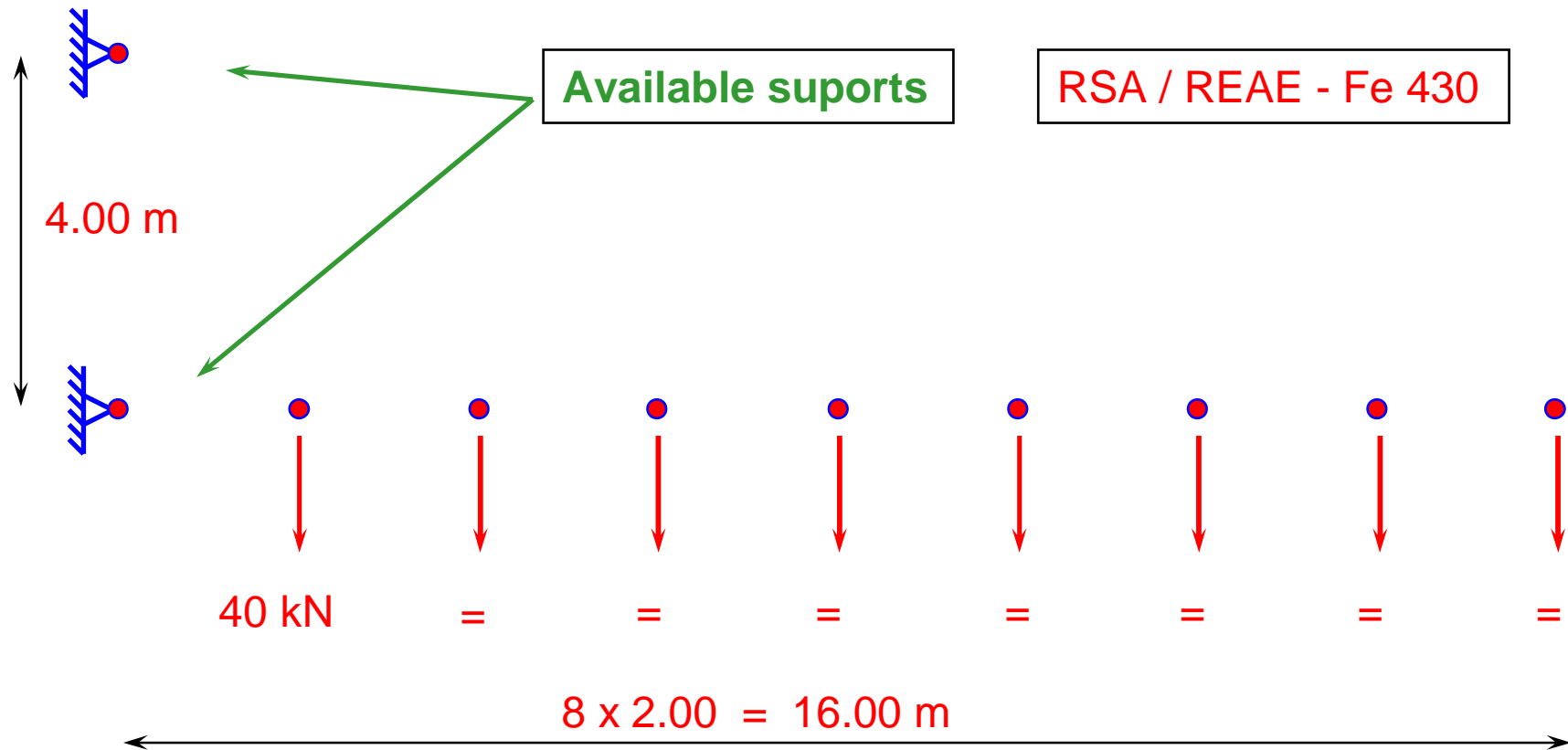


NUMERICAL RESULTS

- Optimal solution - sizing only
 - ◆ Volume = 170 dm³
- Optimal solution - sizing and shape optimization
 - ◆ Volume = 146 dm³ (14 % smaller)
 - ◆ CPU time (PC): less than 10 seconds

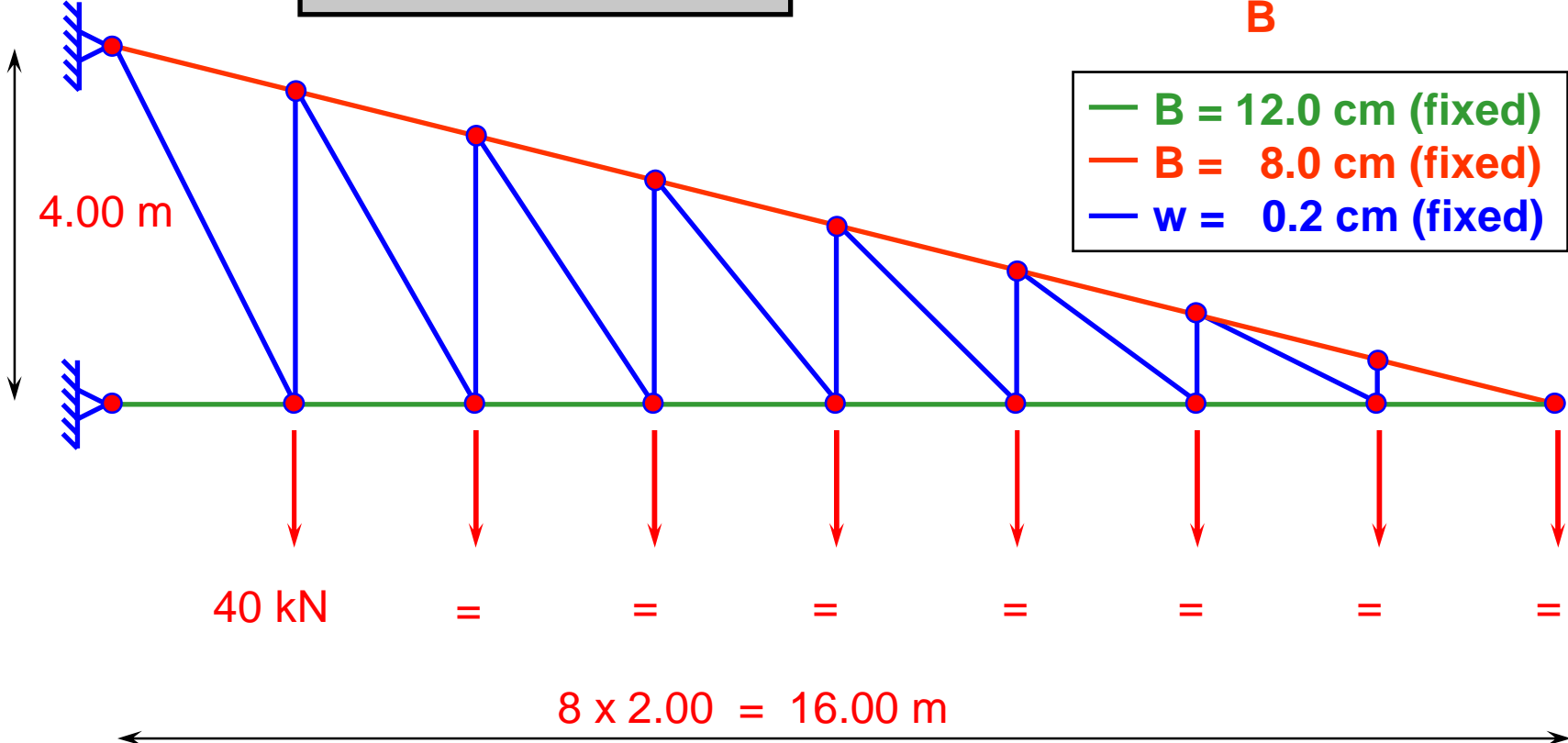
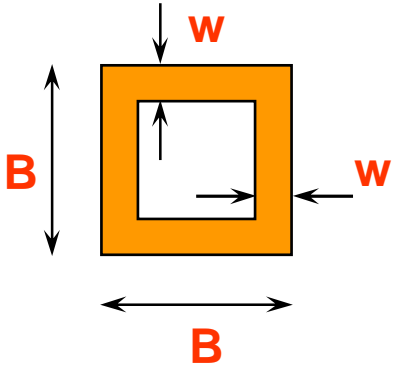
PROBLEM:

create a structure to hold 8 loads of 40 kN each



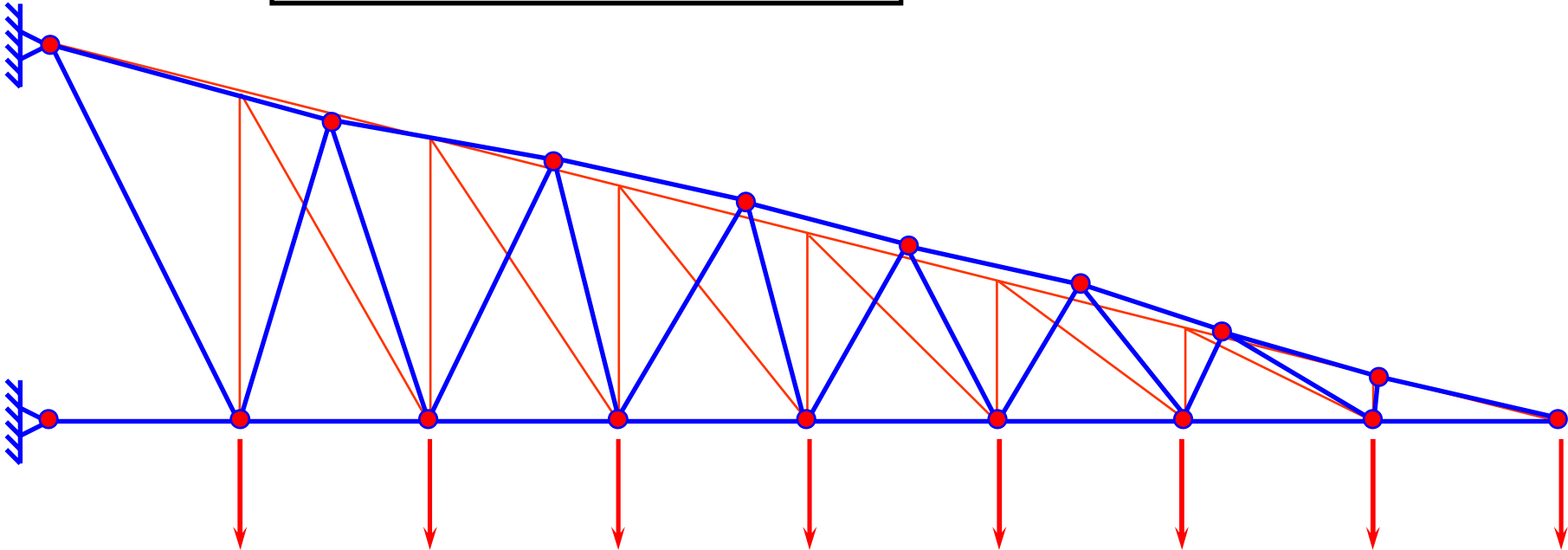
INITIAL SOLUTION

Volume = 69 440 cm³



OPTIMAL SOLUTION

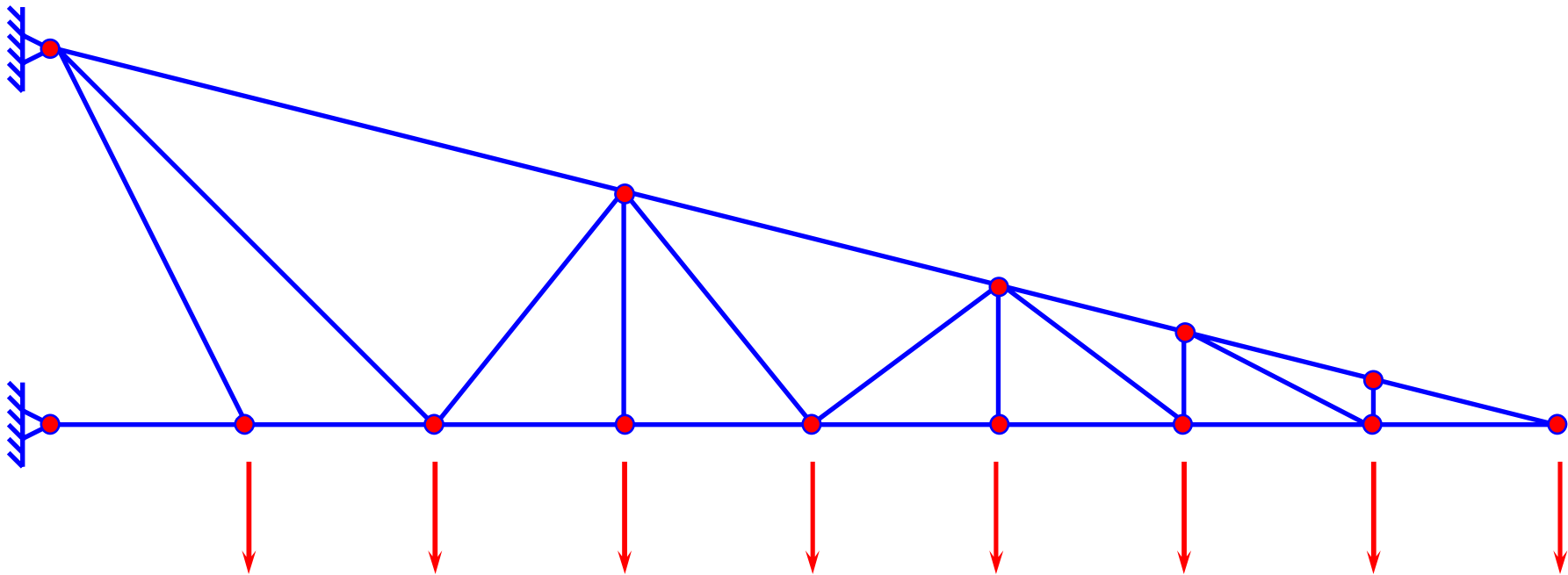
Volume = 66 674 cm³ (- 4 %)



— Initial solution
— Optimal solution

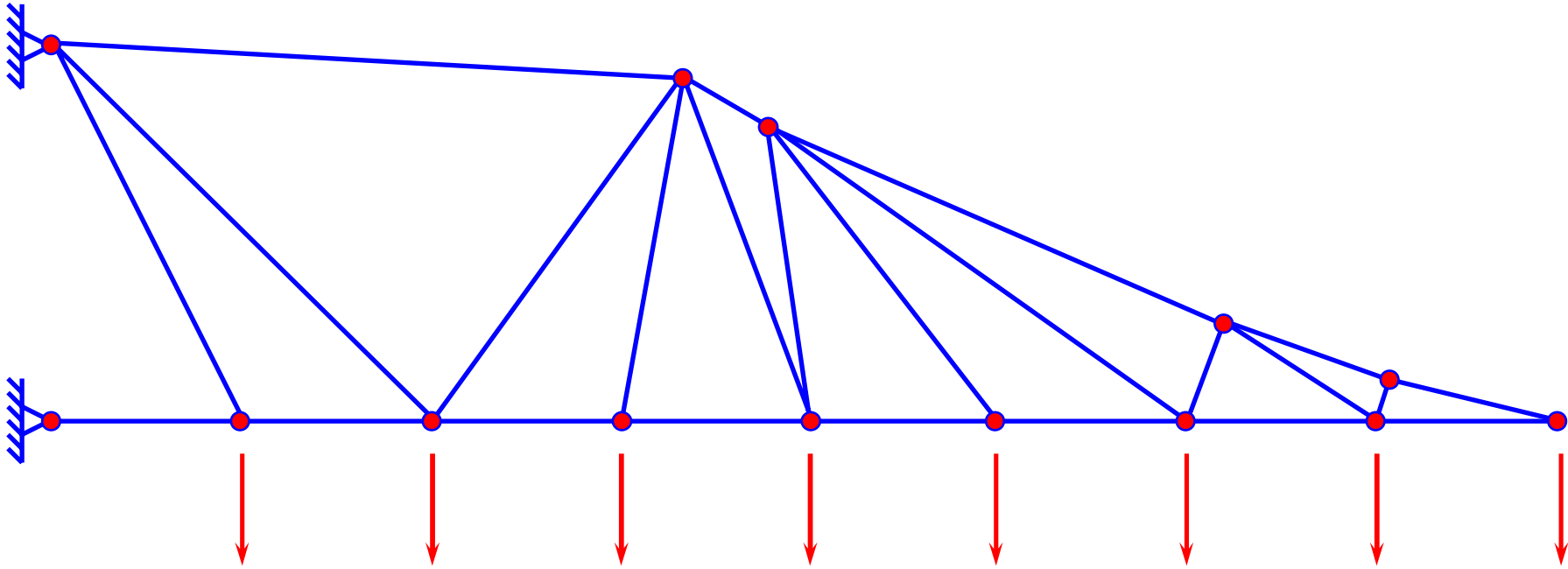
NEW INITIAL SOLUTION

- Same problem
- Distinct topology

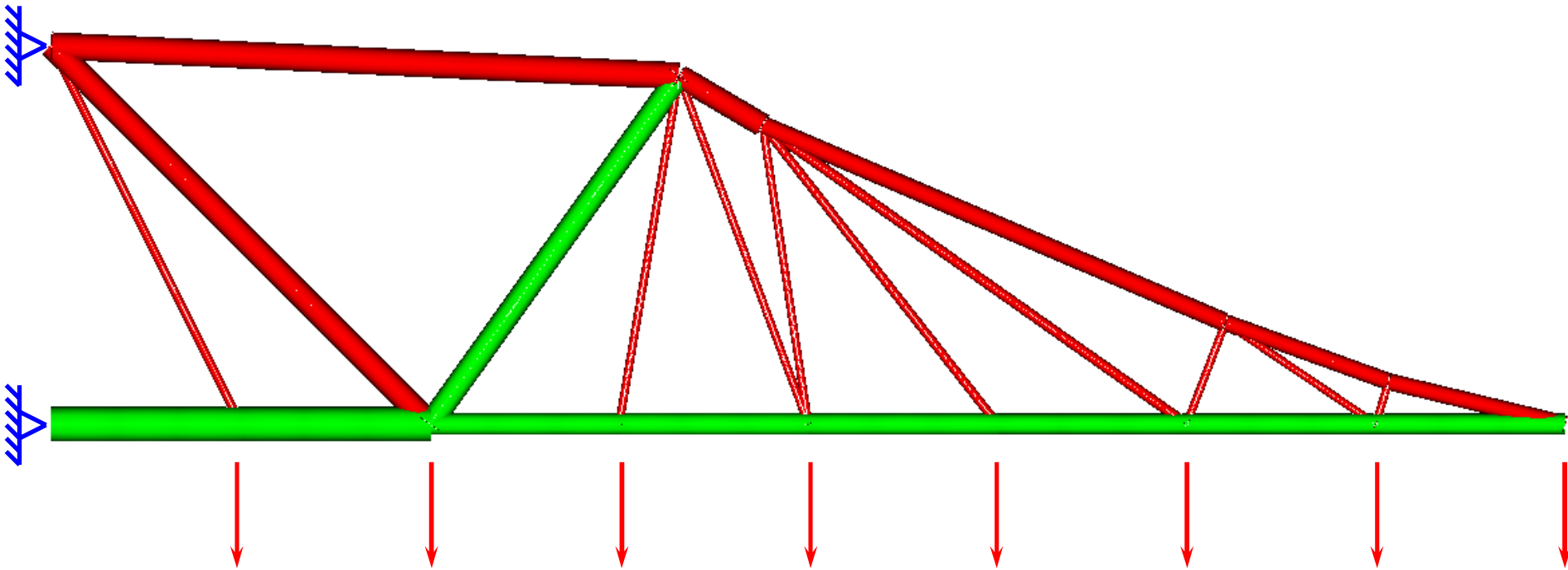


OPTIMAL SOLUTION






Volume = 58 934 cm³ (- 15 %)



OPTIMAL SOLUTION



CONCLUSIONS

-  • Applicable to large scale optimization problems
-  • Very accurate and efficient
-  • Can be used in realistic truss optimization problems
-  • A large number of behavior variables and/or load cases reduces efficiency
-  • Friendly user interface is still required