

# 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

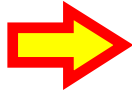
# NONLINEAR PROGRAMMING

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 ;

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- 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

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

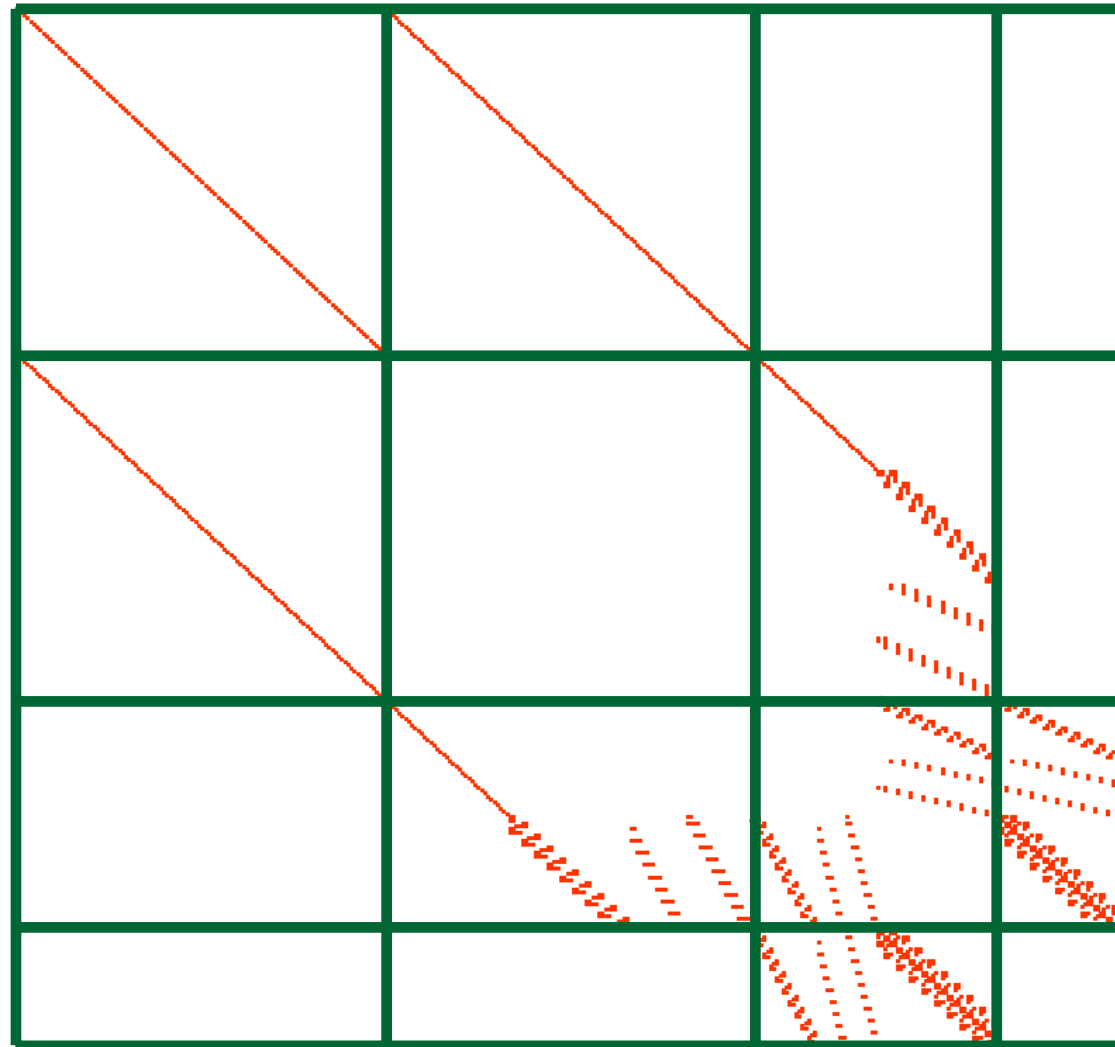
# HESSIAN MATRIX

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

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

# 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  $\alpha$  minimizes the error in  $\Delta \tilde{X}^q$  direction
  - ♦ the value of  $\alpha$  is often close to one
  - ♦ faster convergence
  - ♦ process may fail
- When the value of  $\alpha$  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

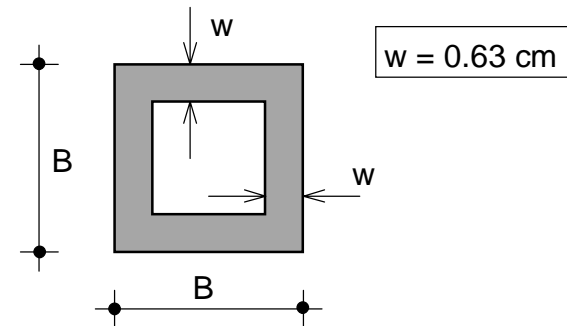


Simultaneously



# 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

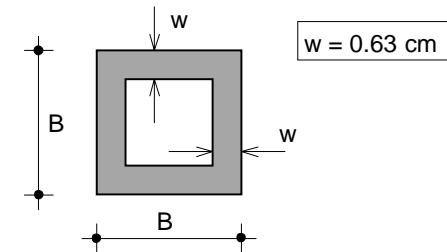
- In most cases the area (A) and the moment of inertia (I) depend on a single parameter (B)

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

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

◆ 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

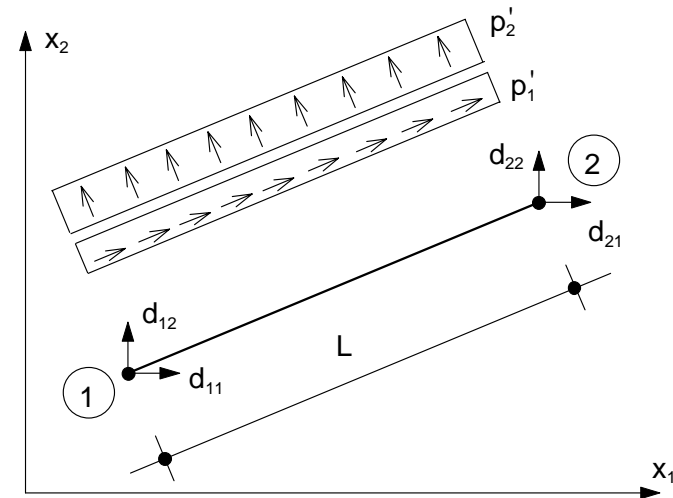


# ADDITIONAL VARIABLES

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

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

◆ Additional variables  $\Rightarrow L_i$



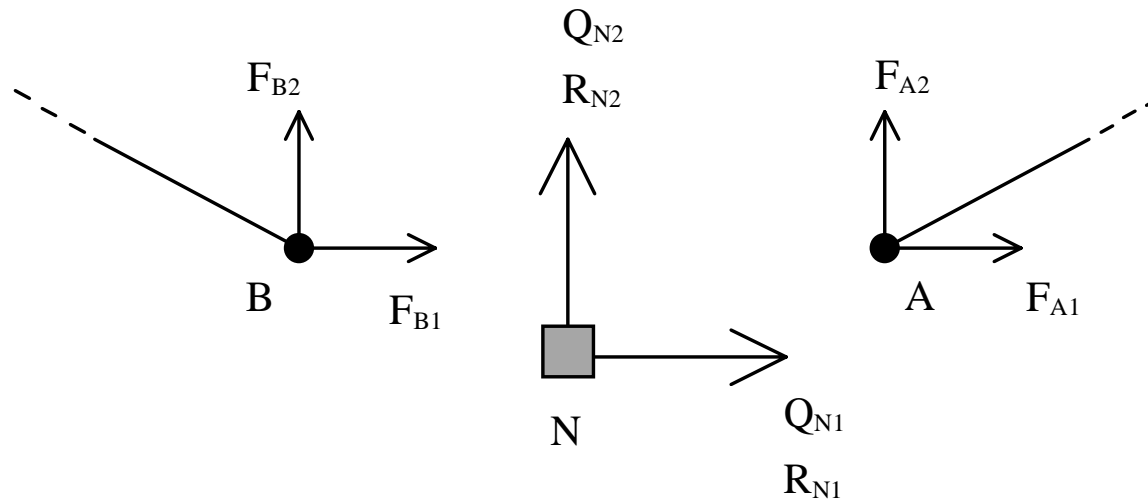
$$-L^2 + x_{11}^2 + x_{12}^2 + x_{21}^2 + x_{22}^2 - 2x_{11}x_{21} - 2x_{12}x_{22} = 0$$

◆ Additional equality constraints  $\Rightarrow L_i$  definition

# EQUILIBRIUM EQUATIONS

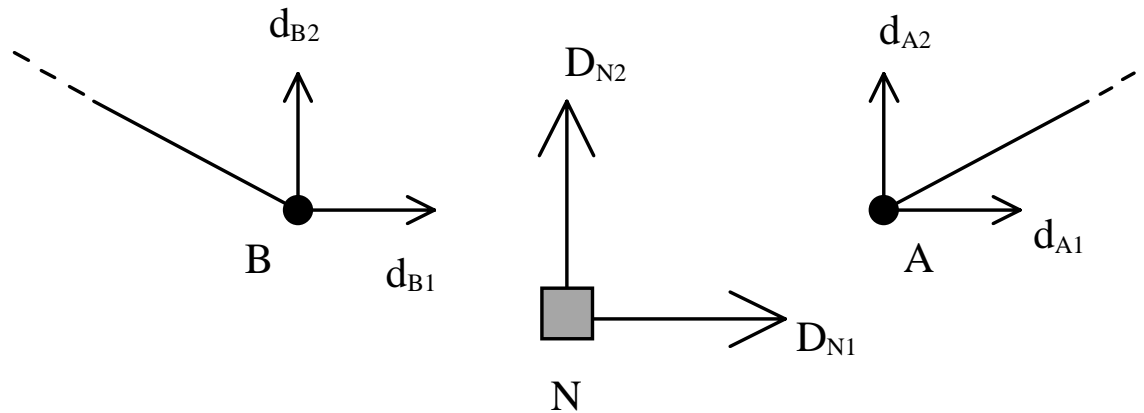
- Equality constraints:

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



- 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:

→ one equation defining L

♦ for each non-prescribed degree of freedom:

→ one equilibrium equation

- Inequality constraints:

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

- ◆ allowable stress (tension and compression)

- ◆ local Euler buckling

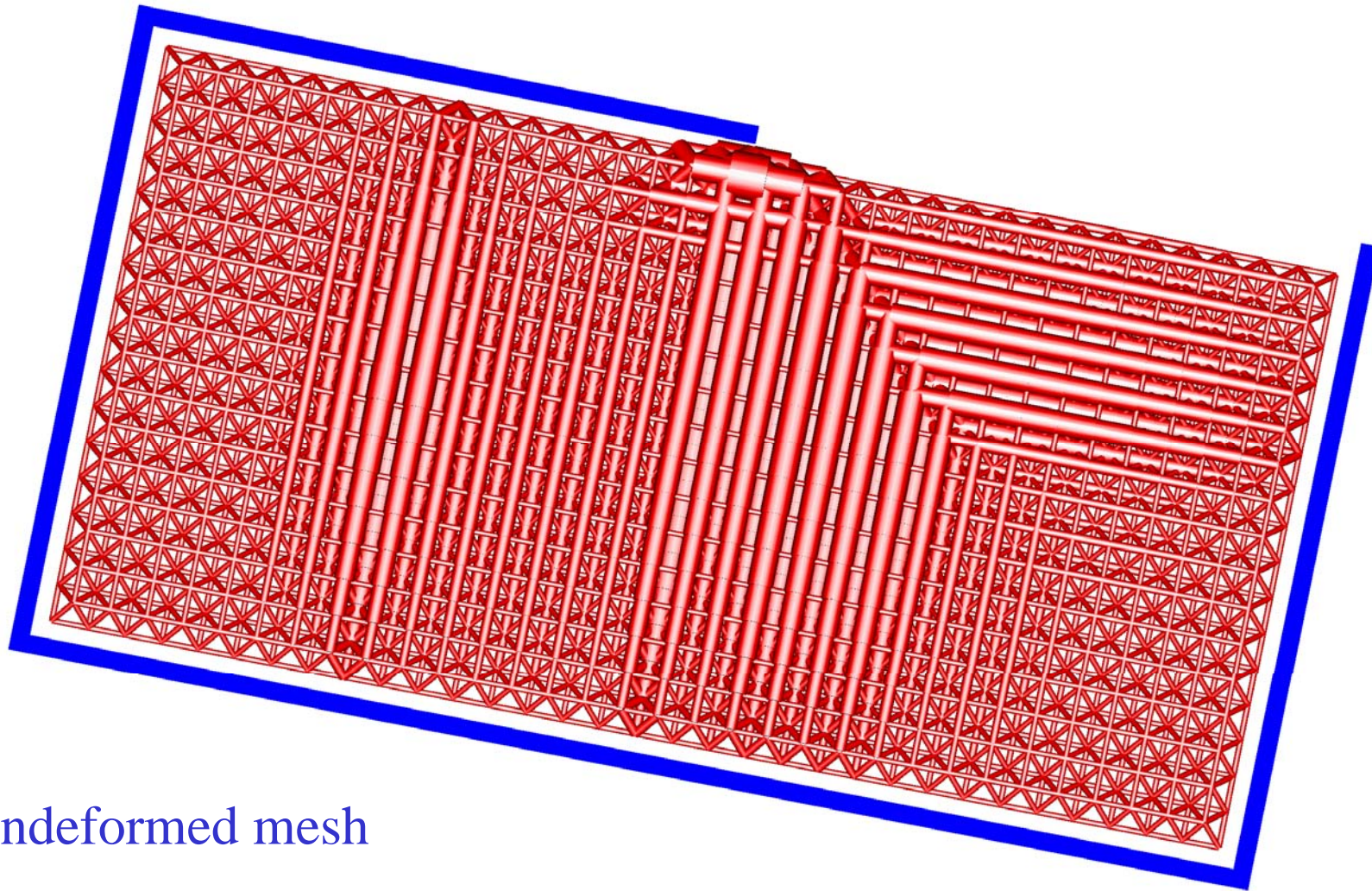
- ◆ side constraints in nodal coordinates  $\Rightarrow x_{\min} \leq x_i \leq x_{\max}$

# 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

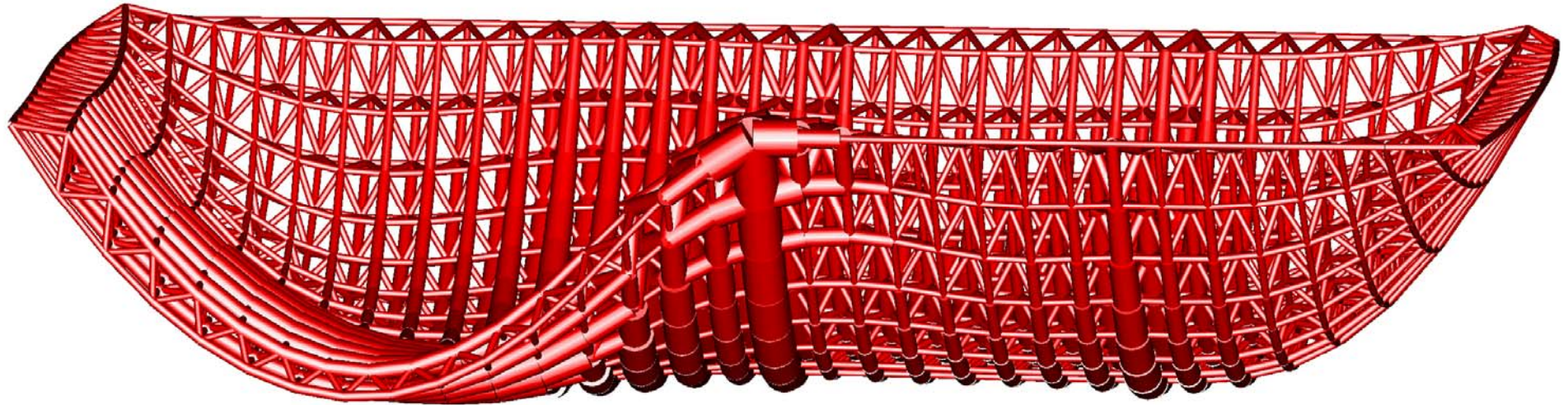


# BUILDING ROOF - OPTIMAL SOLUTION



Undeformed mesh

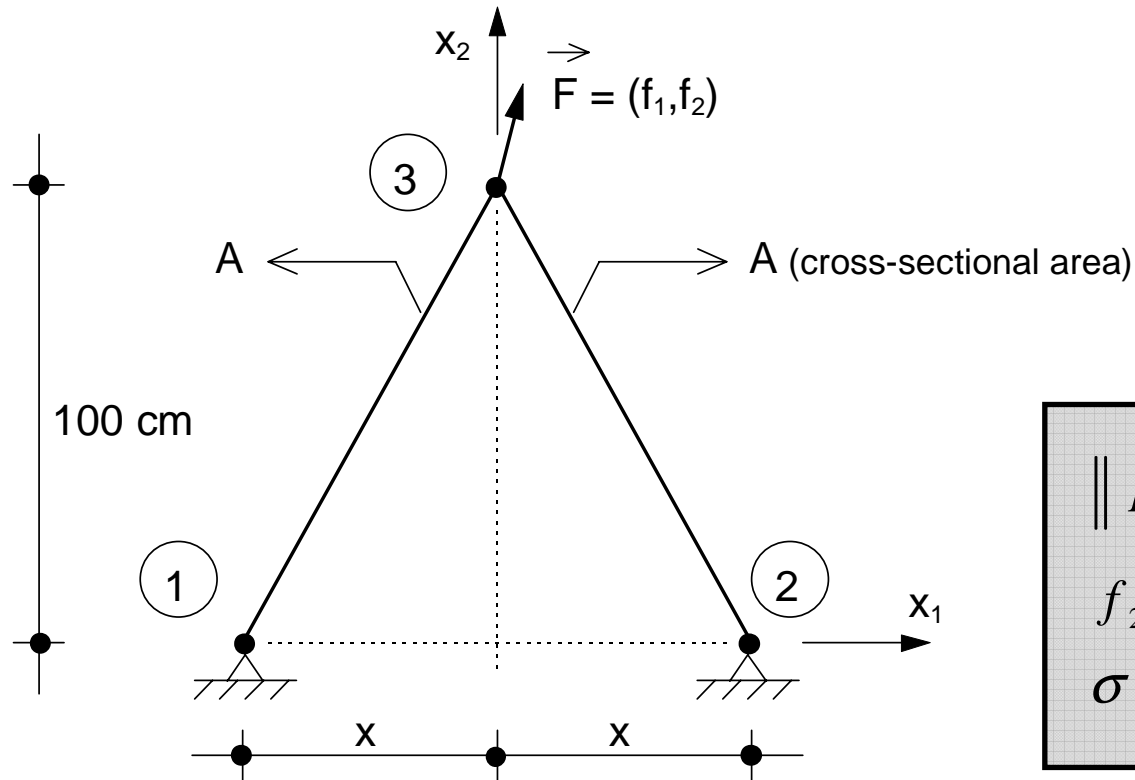
# BUILDING ROOF - OPTIMAL SOLUTION



Deformed mesh



# SHAPE OPTIMIZATION TEST PROBLEM



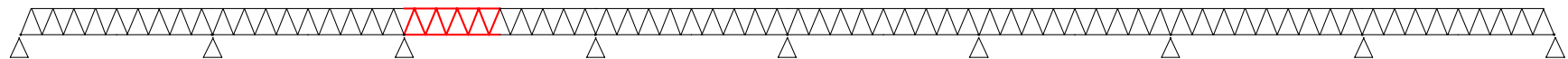
$$\begin{aligned} \|\vec{F}\| &= 200 \text{ kN} \\ f_2 &= 8 f_1 \\ \sigma_{\max} &= 100 \text{ kN/cm}^2 \end{aligned}$$

• Variables:  $A, x$

• Svanberg's solution confirmed

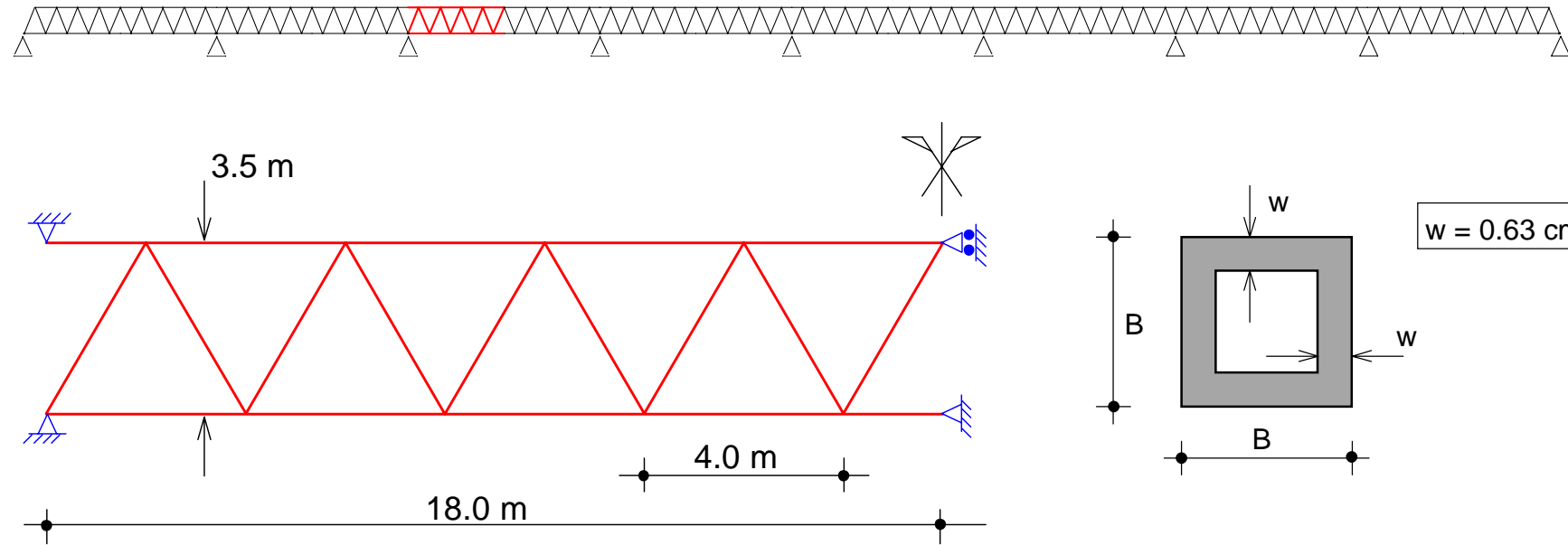
# SHAPE OPTIMIZATION PROBLEM

- 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



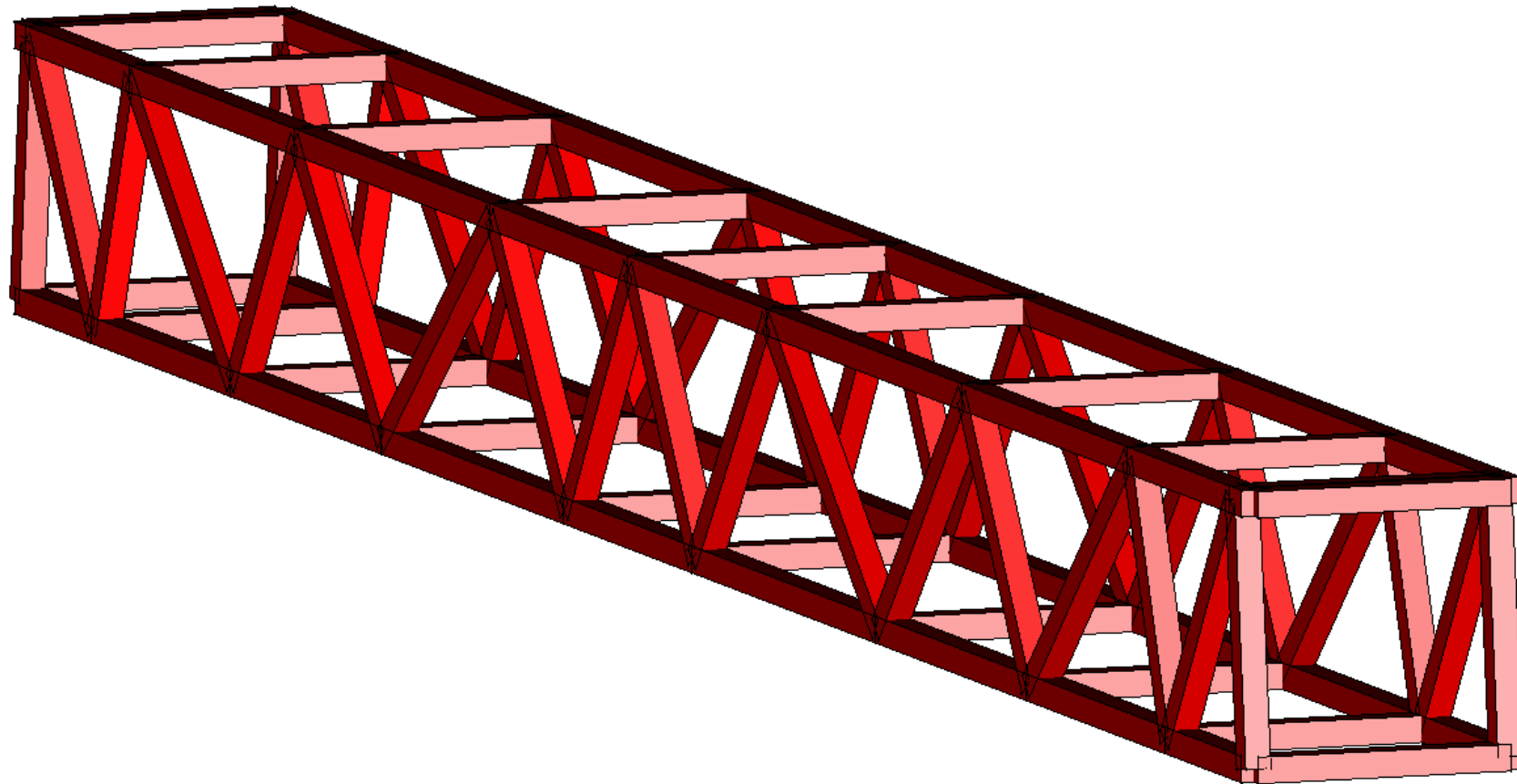
# STEEL BRIDGE

Vertical distributed load

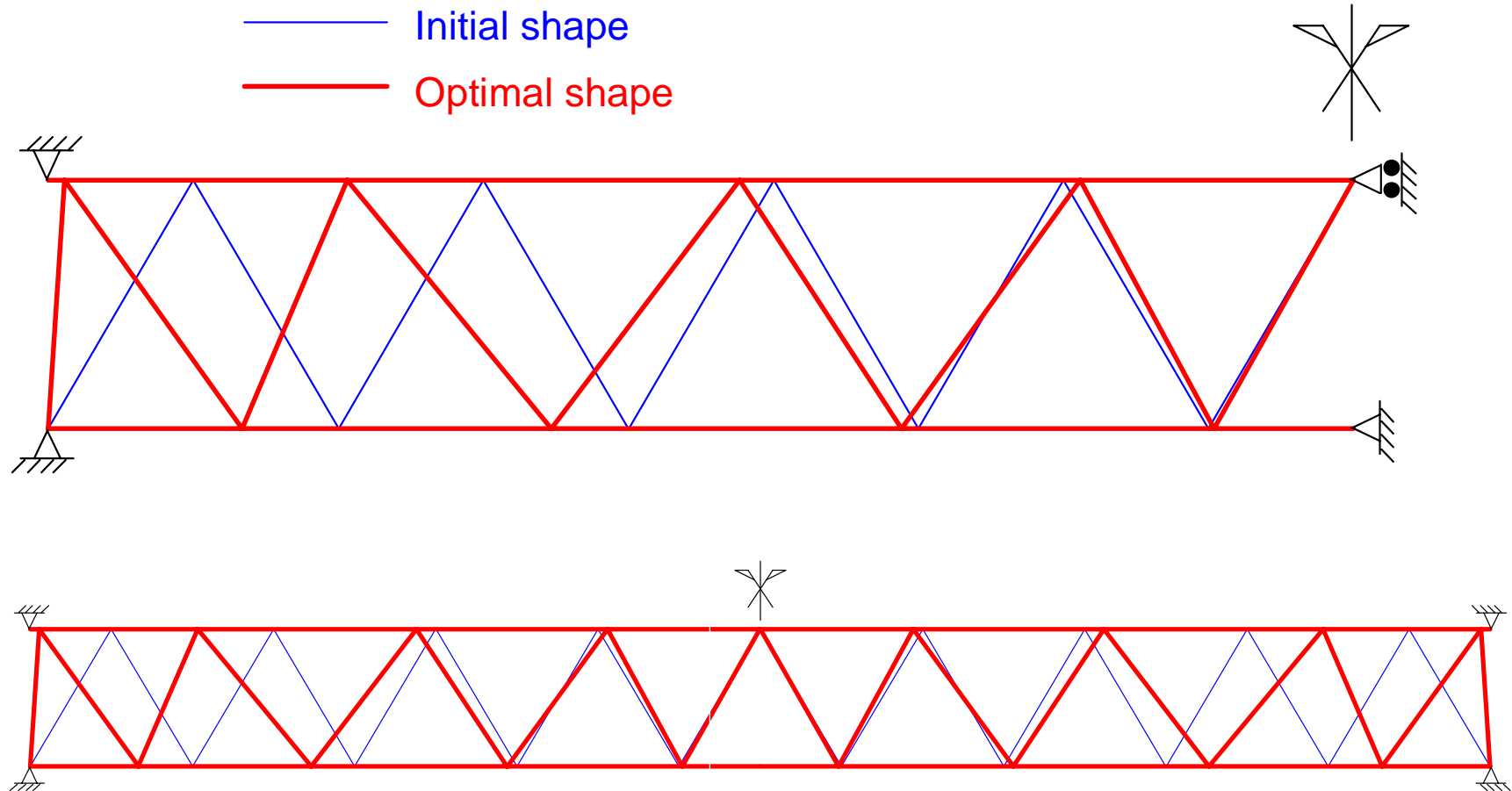


- Group I - horizontal bars
- Group II - diagonal bars

# STEEL BRIDGE



# OPTIMAL SHAPE



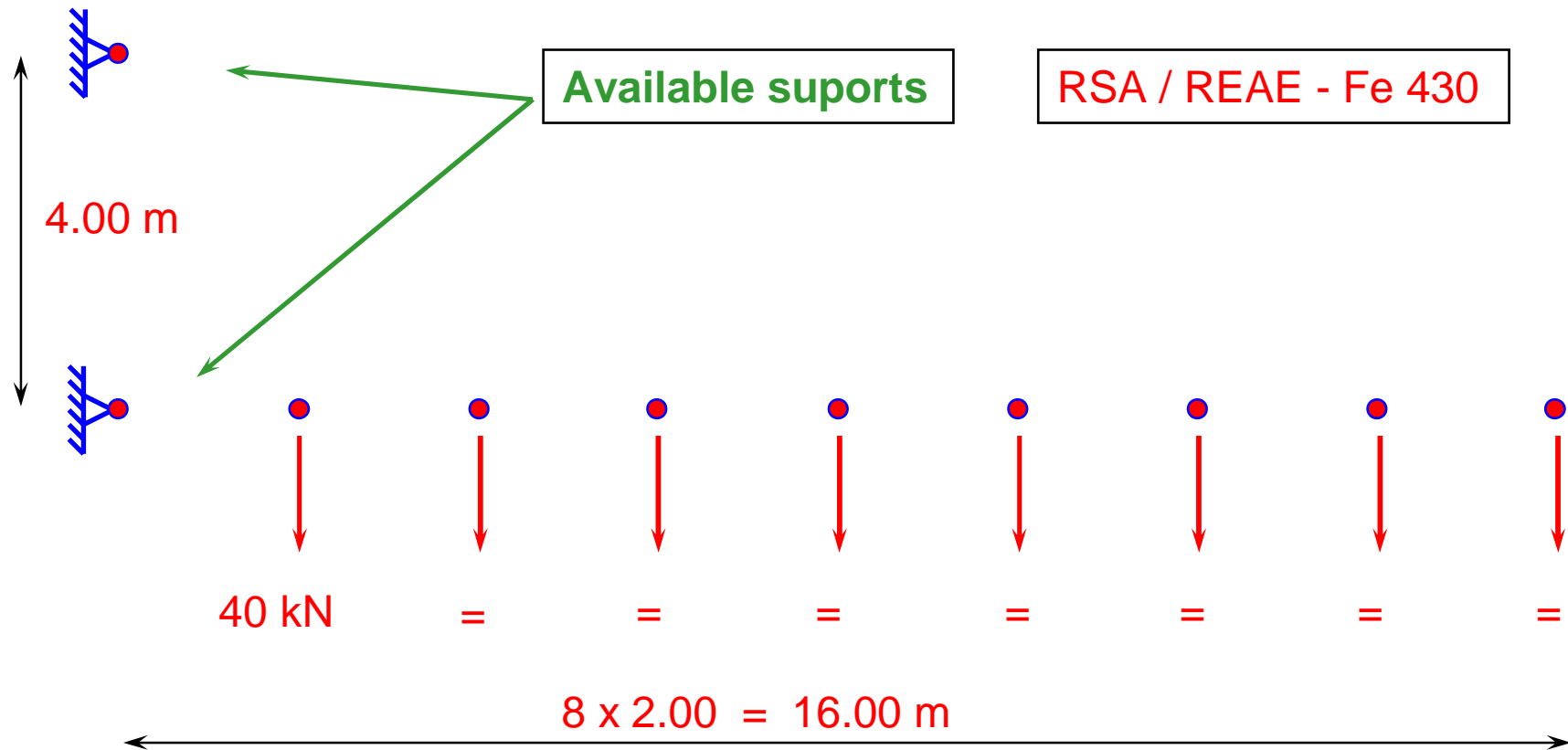
# NUMERICAL RESULTS

- Optimal solution - sizing only
  - ♦ Volume = 170 dm<sup>3</sup>
- Optimal solution - sizing and shape optimization
  - ♦ Volume = 146 dm<sup>3</sup> (14 % smaller)
  - ♦ CPU time (PC): less than 10 seconds



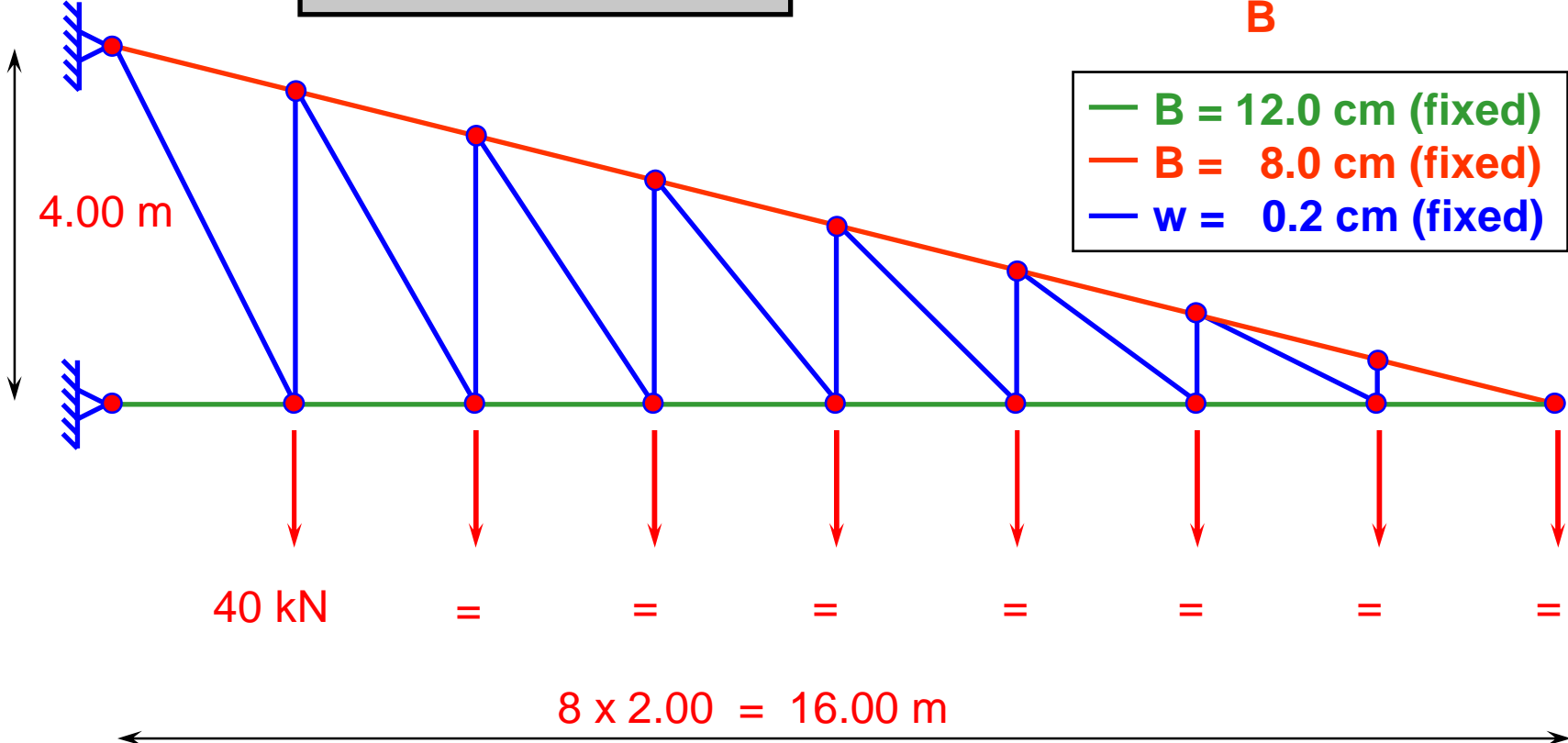
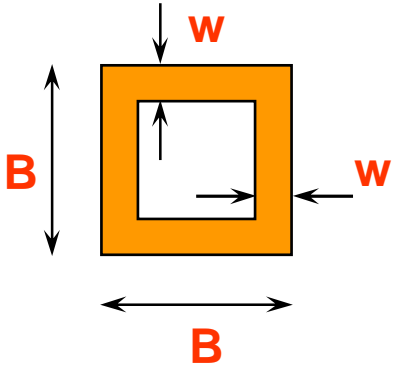
# PROBLEM:

create a structure to hold 8 loads of 40 kN each



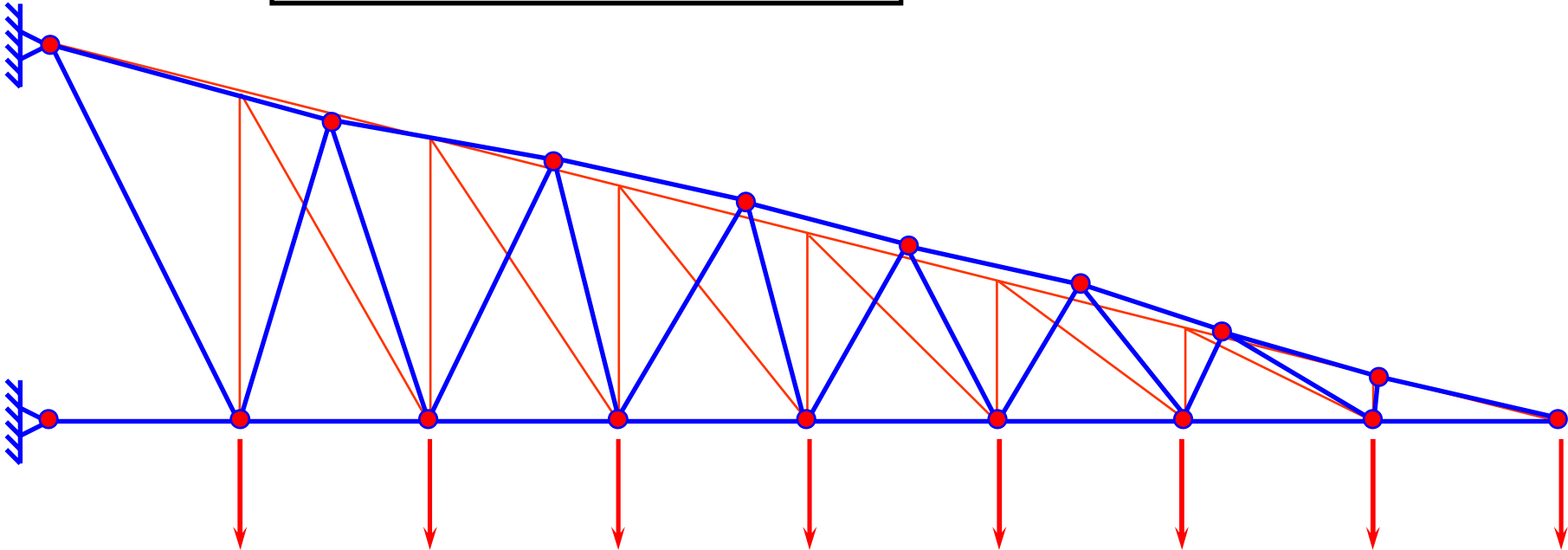
# INITIAL SOLUTION

Volume = 69 440 cm<sup>3</sup>



# OPTIMAL SOLUTION

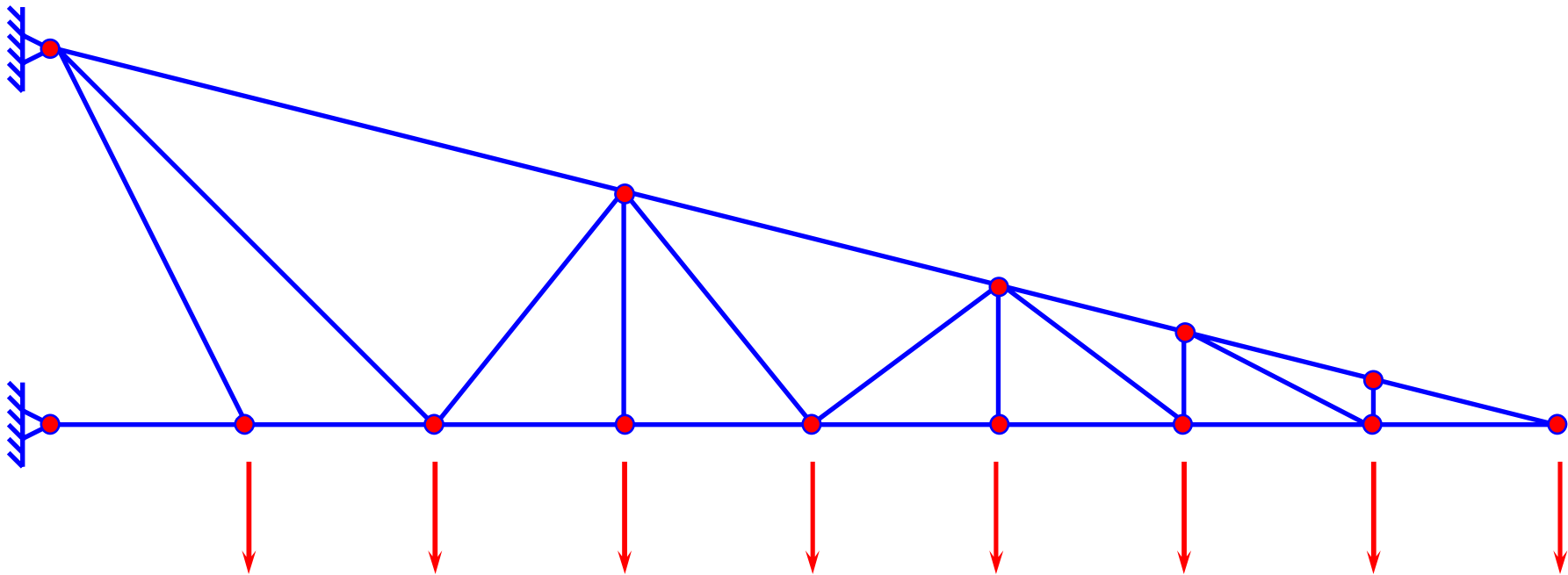
Volume = 66 674 cm<sup>3</sup> ( - 4 % )



— Initial solution  
— Optimal solution

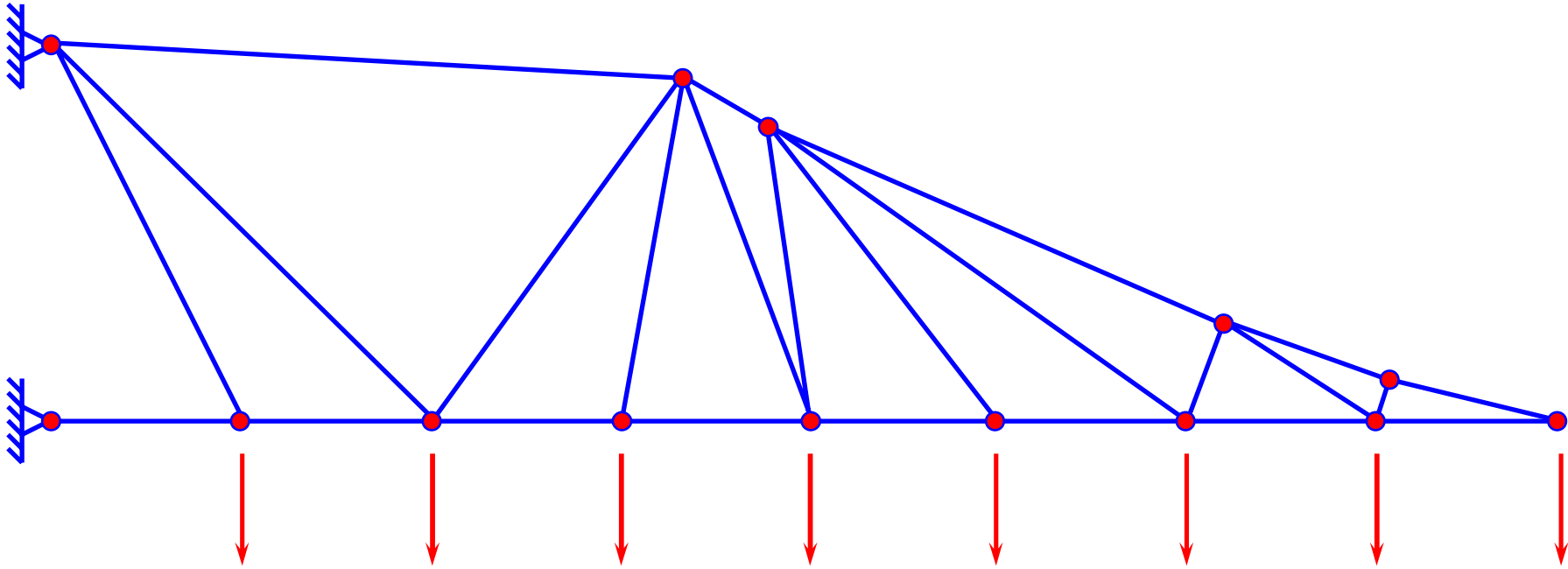
# NEW INITIAL SOLUTION

- Same problem
- Distinct topology

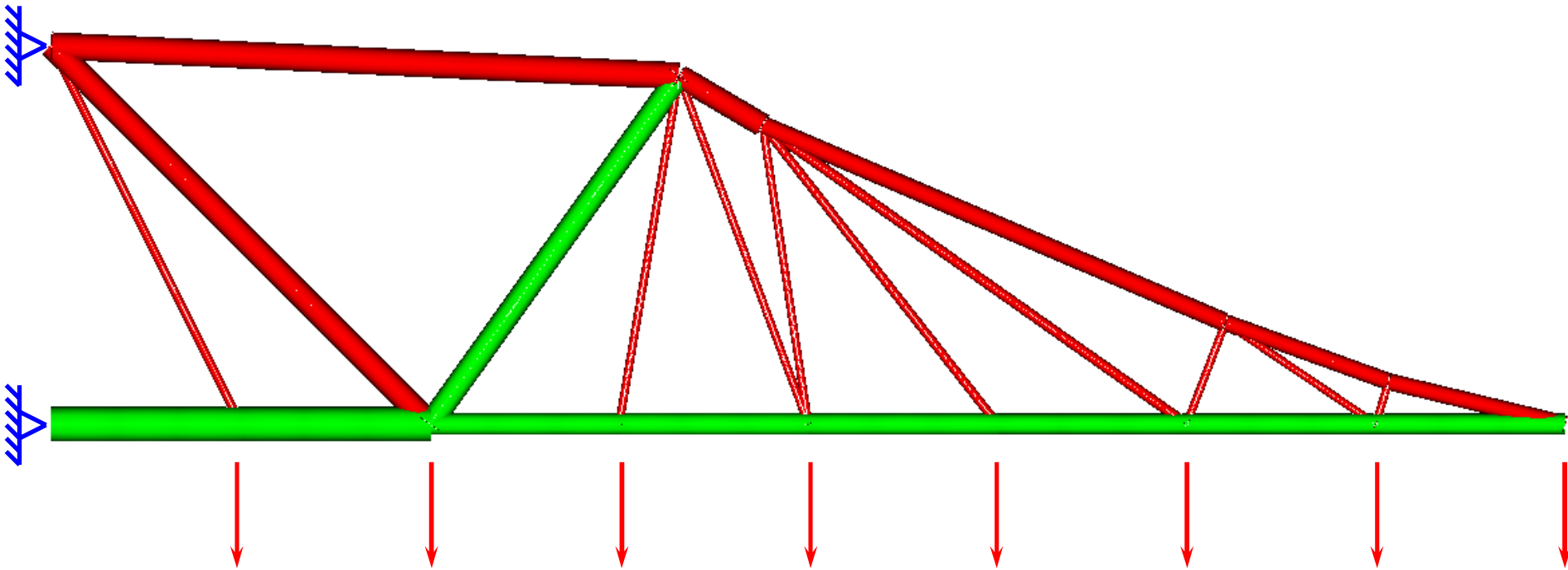


# OPTIMAL SOLUTION






Volume = 58 934 cm<sup>3</sup> ( - 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