

# Geometric Visualization of LP Algorithms

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# Structural Optimization

**Forces:**  $x_{ij}$  = tension in *beam* (aka *member*)  $\{i, j\}$ .

- $x_{ij} = x_{ji}$ .
- Compression = -Tension.

**Force Balance:**

Look at joint 2:

$$x_{12} \begin{bmatrix} -1 \\ 0 \end{bmatrix} + x_{23} \begin{bmatrix} -0.6 \\ 0.8 \end{bmatrix} + x_{24} \begin{bmatrix} 0 \\ 1 \end{bmatrix} = - \begin{bmatrix} b_2^1 \\ b_2^2 \end{bmatrix}$$

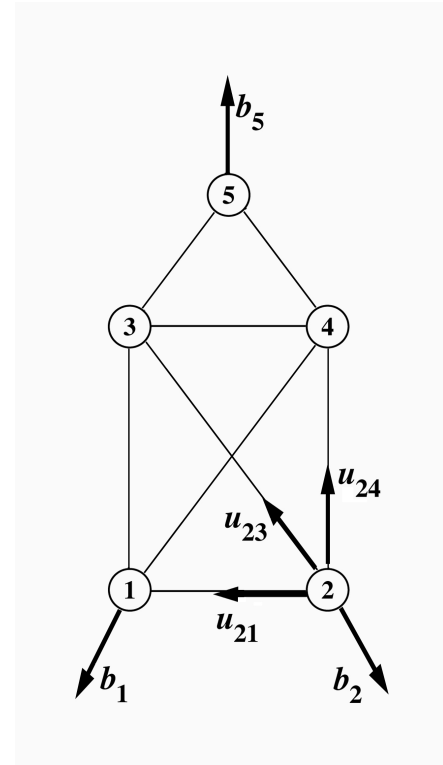
**Notations:**

$\vec{p}_i$  = position vector for joint  $i$

$$\vec{u}_{ij} = \frac{\vec{p}_j - \vec{p}_i}{\|\vec{p}_j - \vec{p}_i\|} \quad (\text{Note } \vec{u}_{ji} = -\vec{u}_{ij})$$

**Constraints:**

$$\sum_{\substack{j: \\ \{i,j\} \in \mathcal{A}}} \vec{u}_{ij} x_{ij} = -\vec{b}_i \quad i = 1, \dots, m.$$



# Matrix Form

$$Ax = -b$$

$$x^T = [ \quad x_{12} \quad x_{13} \quad x_{14} \quad x_{23} \quad x_{24} \quad x_{34} \quad x_{35} \quad x_{45} \quad ]$$

$$A = \begin{matrix} & \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} & \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} & \begin{bmatrix} .6 \\ .8 \end{bmatrix} & & & & & & \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & & \begin{bmatrix} -1 \\ 0 \end{bmatrix} & & & \begin{bmatrix} -.6 \\ .8 \end{bmatrix} & \begin{bmatrix} 0 \\ 1 \end{bmatrix} & & & & \\ & & & \begin{bmatrix} 0 \\ -1 \end{bmatrix} & & \begin{bmatrix} .6 \\ -.8 \end{bmatrix} & & \begin{bmatrix} 1 \\ 0 \end{bmatrix} & \begin{bmatrix} .6 \\ .8 \end{bmatrix} & & \\ & & & & \begin{bmatrix} -.6 \\ -.8 \end{bmatrix} & & \begin{bmatrix} 0 \\ -1 \end{bmatrix} & \begin{bmatrix} -1 \\ 0 \end{bmatrix} & & \begin{bmatrix} -.6 \\ .8 \end{bmatrix} & \\ & & & & & & & & \begin{bmatrix} -.6 \\ -.8 \end{bmatrix} & \begin{bmatrix} .6 \\ -.8 \end{bmatrix} & \end{matrix}, \quad b = \begin{bmatrix} b_1^1 \\ b_1^2 \\ b_2^1 \\ b_2^2 \\ b_3^1 \\ b_3^2 \\ b_4^1 \\ b_4^2 \\ b_5^1 \\ b_5^2 \end{bmatrix}.$$

Notes:

- $\|\vec{u}_{ij}\| = \|\vec{u}_{ji}\| = 1$ .
- $\vec{u}_{ij} = -\vec{u}_{ji}$ .
- Each column contains a  $\vec{u}_{ij}$ , a  $\vec{u}_{ji}$ , and rest are zero.
- In one dimension, exactly a node-arc incidence matrix.

# Minimum Weight Structural Design

$$\begin{aligned} & \text{minimize} && \sum_{\{i,j\} \in \mathcal{A}} l_{ij} |x_{ij}| \\ & \text{subject to} && \sum_{\substack{j: \\ \{i,j\} \in \mathcal{A}}} \vec{u}_{ij} x_{ij} = -\vec{b}_i \quad i = 1, 2, \dots, m. \end{aligned}$$

Not quite an LP.

Use a common trick:

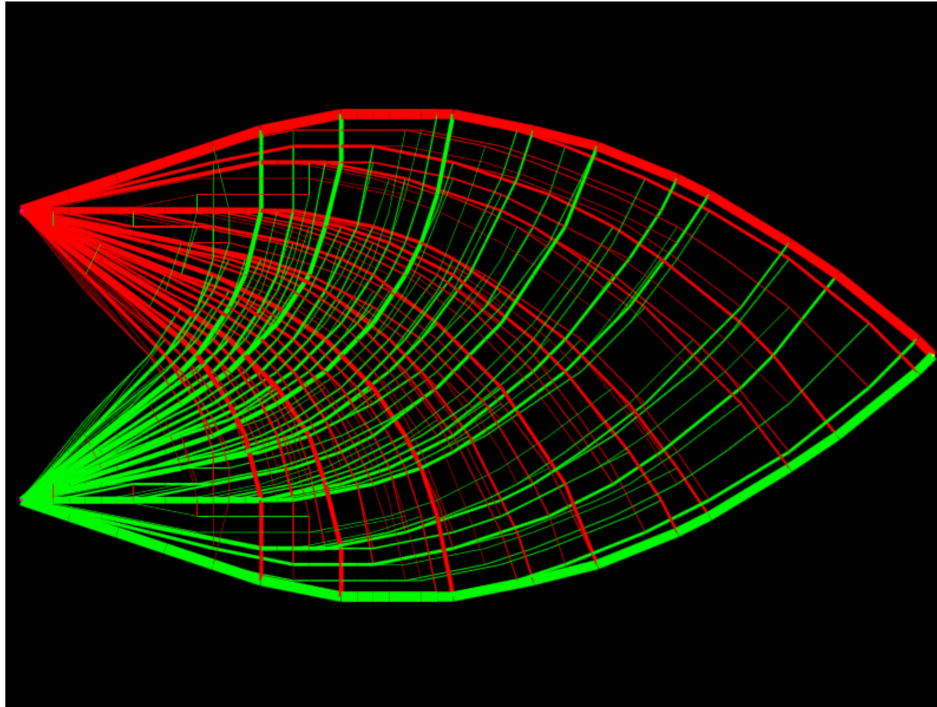
$$\begin{aligned} x_{ij} &= x_{ij}^+ - x_{ij}^-, & x_{ij}^+, x_{ij}^- &\geq 0 \\ |x_{ij}| &= x_{ij}^+ + x_{ij}^- \end{aligned}$$

Reformulated as an LP:

$$\begin{aligned} & \text{minimize} && \sum_{\{i,j\} \in \mathcal{A}} (l_{ij} x_{ij}^+ + l_{ij} x_{ij}^-) \\ & \text{subject to} && \sum_{\substack{j: \\ \{i,j\} \in \mathcal{A}}} (\vec{u}_{ij} x_{ij}^+ - \vec{u}_{ij} x_{ij}^-) = -\vec{b}_i \quad i = 1, 2, \dots, m \\ & && x_{ij}^+, x_{ij}^- \geq 0 \quad \{i, j\} \in \mathcal{A}. \end{aligned}$$

# The Michell Bracket (1904)

Height:  Width:  Step Length:  Beam Strength:  x-load:  y-load:



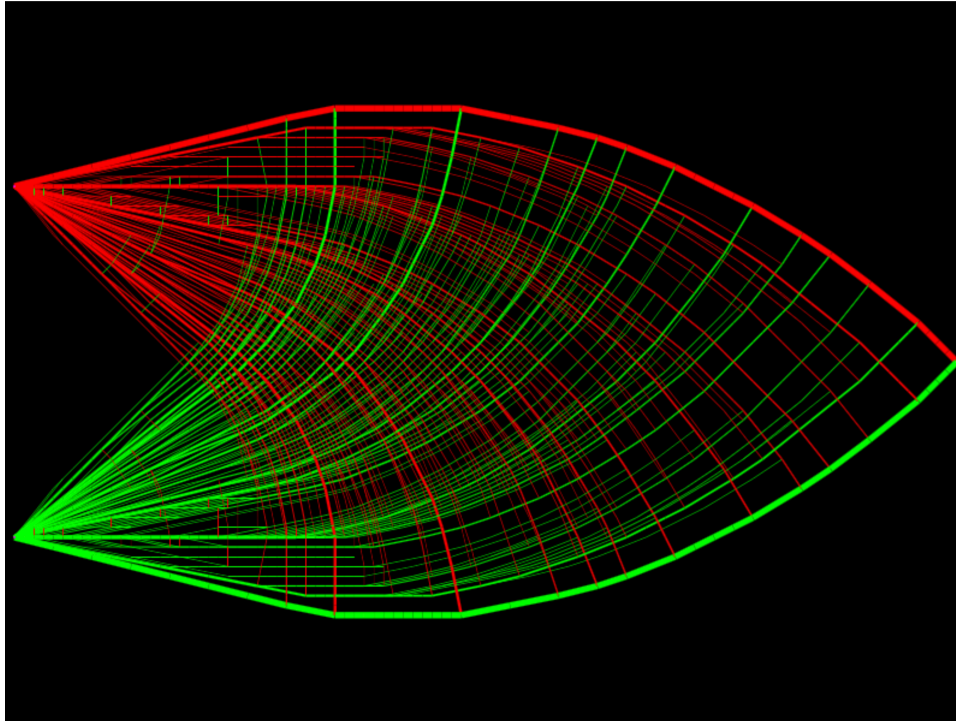
Constraints: 5,396  
Variables: 193,310  
Time: 270 seconds

Click [here](#) for parametric self-dual simplex method animation tool.

Click [here](#) for affine-scaling method animation tool.

# The Michell Bracket (1904)

Height:  Width:  Step Length:  Beam Strength:  x-load:  y-load:



Constraints: 14,996  
Variables: 536,972  
Time: 1994 seconds

Click [here](#) for parametric self-dual simplex method animation tool.

Click [here](#) for affine-scaling method animation tool.

# Algorithm Animation: Two-Phase Simplex Method

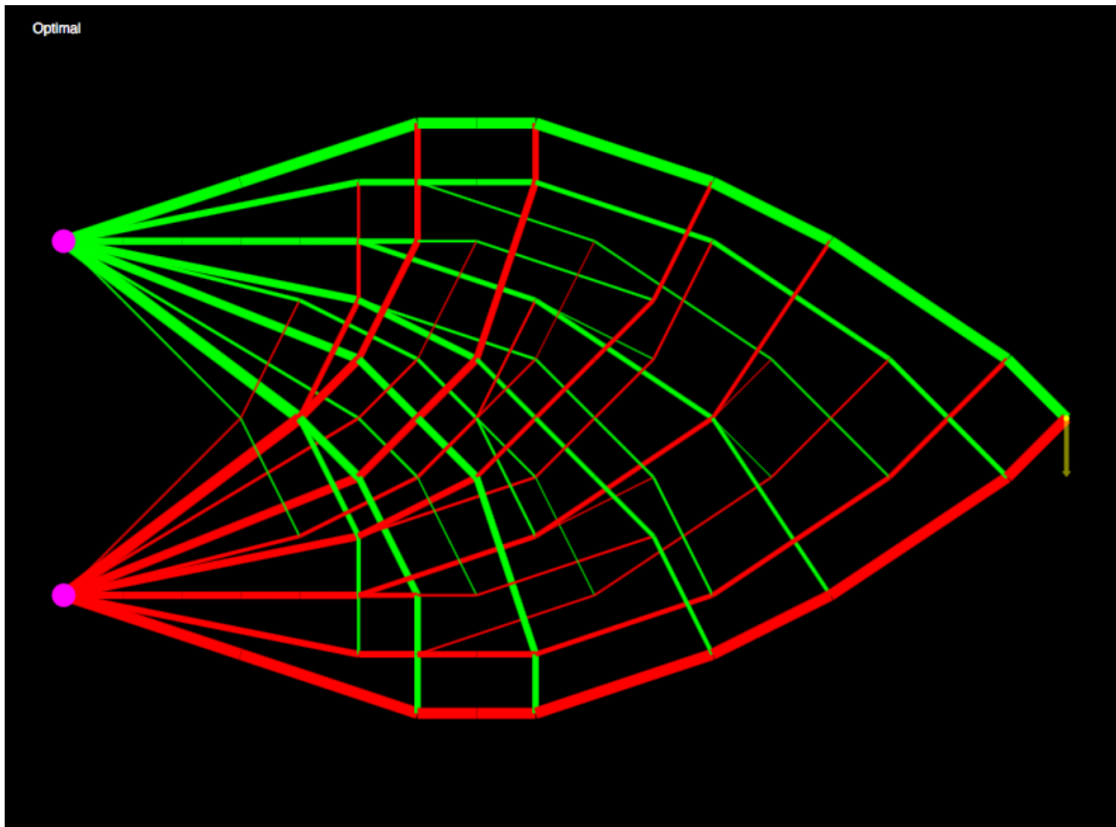
## via Structural Optimization

(Click [here](#) for older Java version.)

Height:  Width:

Use the mouse with x-load:  y-load:  ... or ... Select from this list:

Beam Strength:  Show:



# Algorithm Animation: Primal-Dual Simplex Method

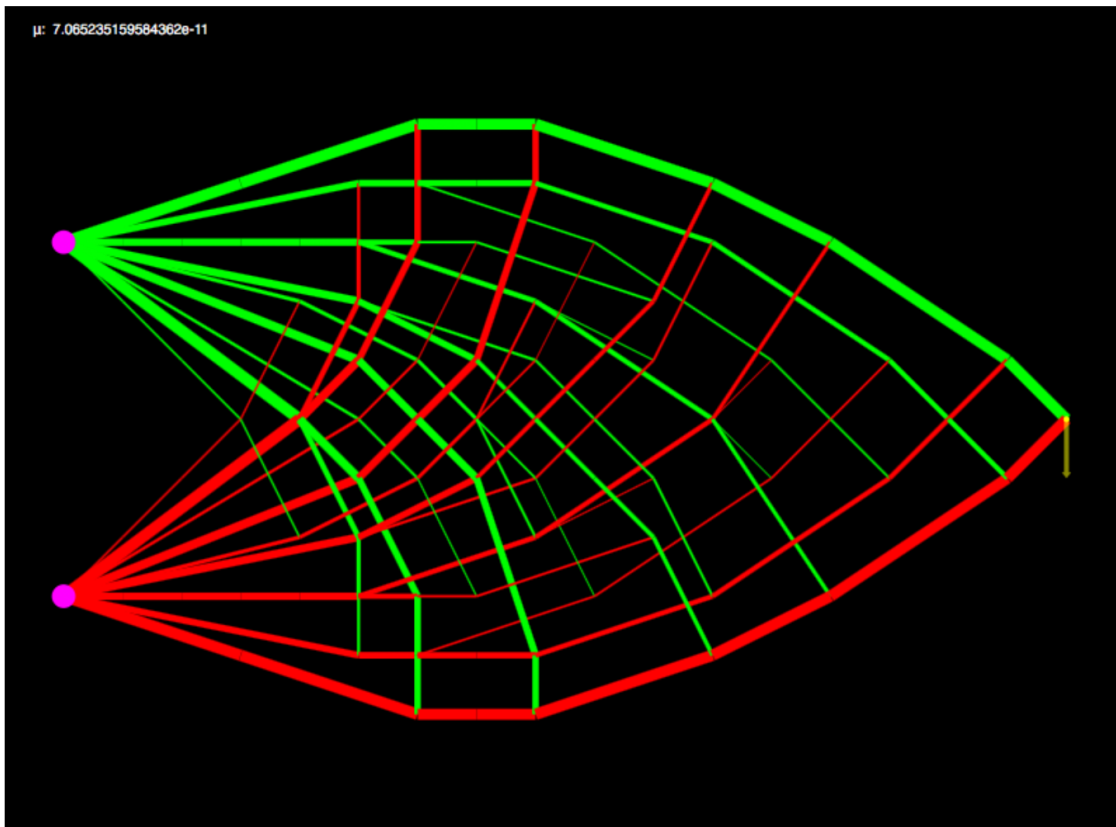
## via Structural Optimization

(Click [here](#) for older Java version.)

Height:  Width:

Use the mouse with x-load:  y-load:  ... or ... Select from this list:

Beam Strength:  Show:



# Algorithm Animation: Affine Scaling

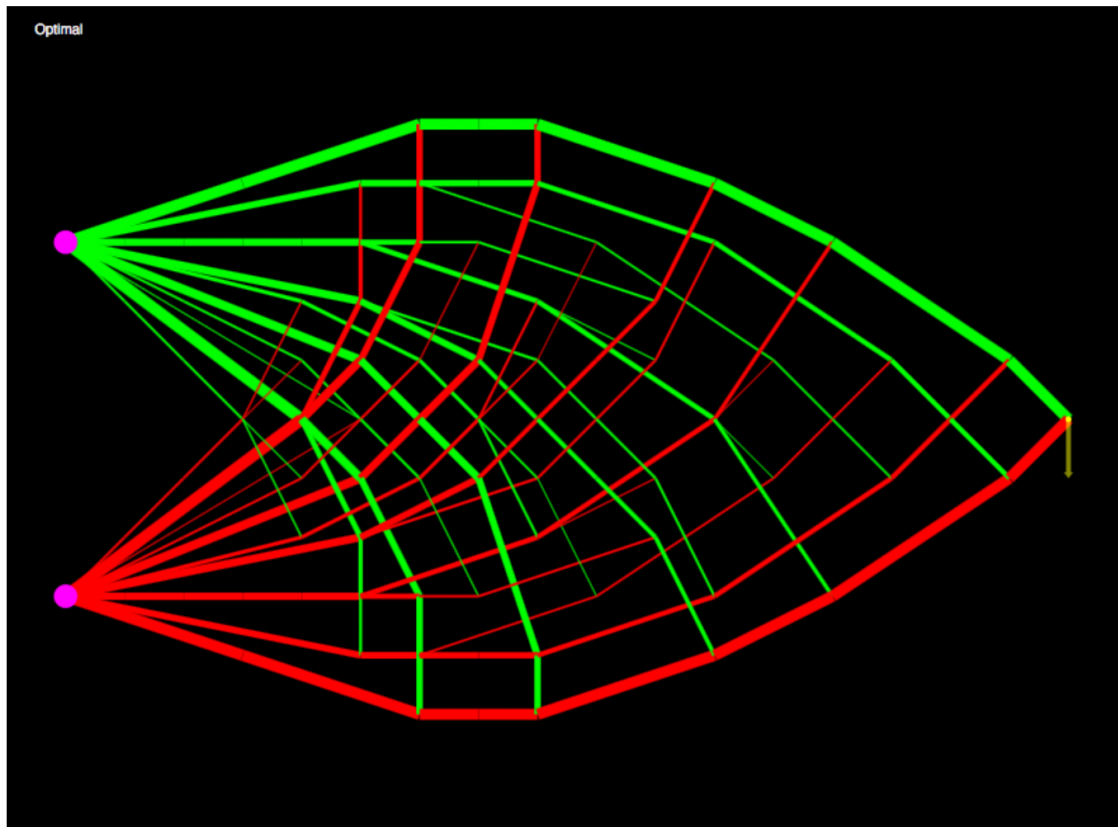
## via Structural Optimization

(Click [here](#) for older Java version.)

Height:  Width:  Neighbors:

Use the mouse with x-load:  y-load:  ... or ... Select from list

Step Length:  Beam Strength:



# Algorithm Animation: Two-Phase Simplex Method

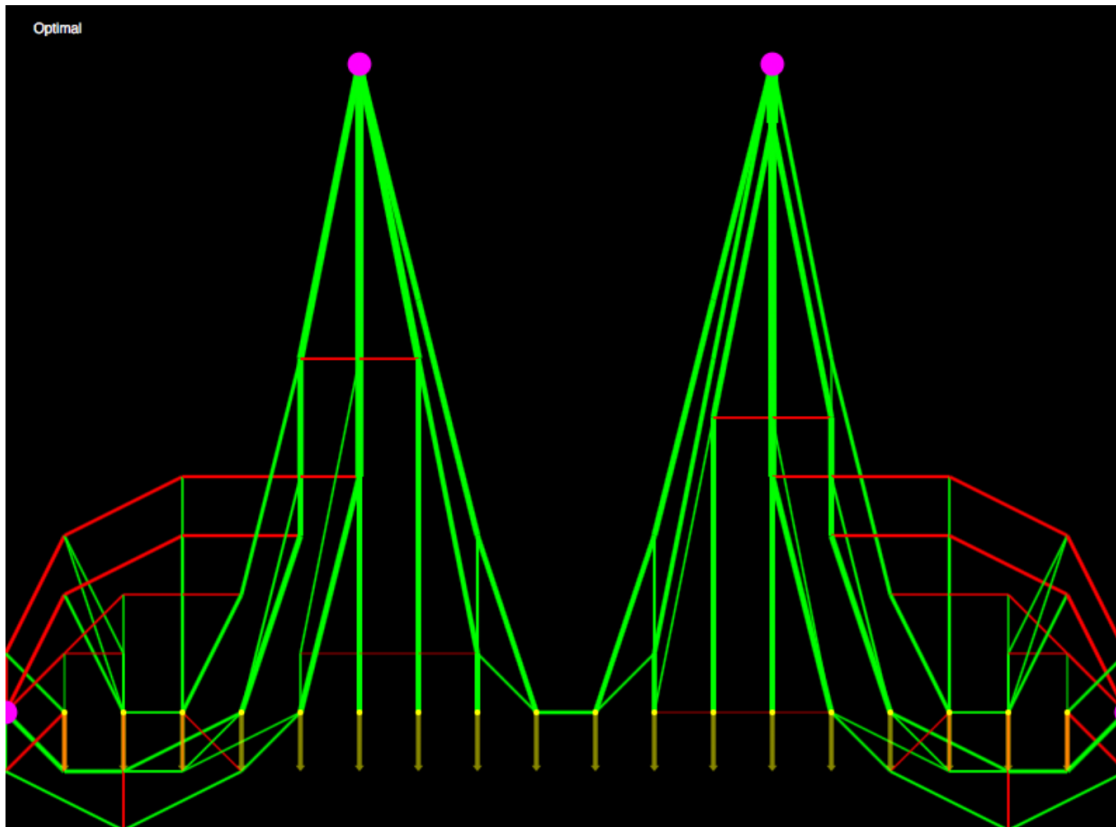
## via Structural Optimization

(Click [here](#) for older Java version.)

Height:  Width:

Use the mouse with x-load:  y-load:  ... or ... Select from this list:

Beam Strength:  Show:



# Algorithm Animation: Primal-Dual Simplex Method

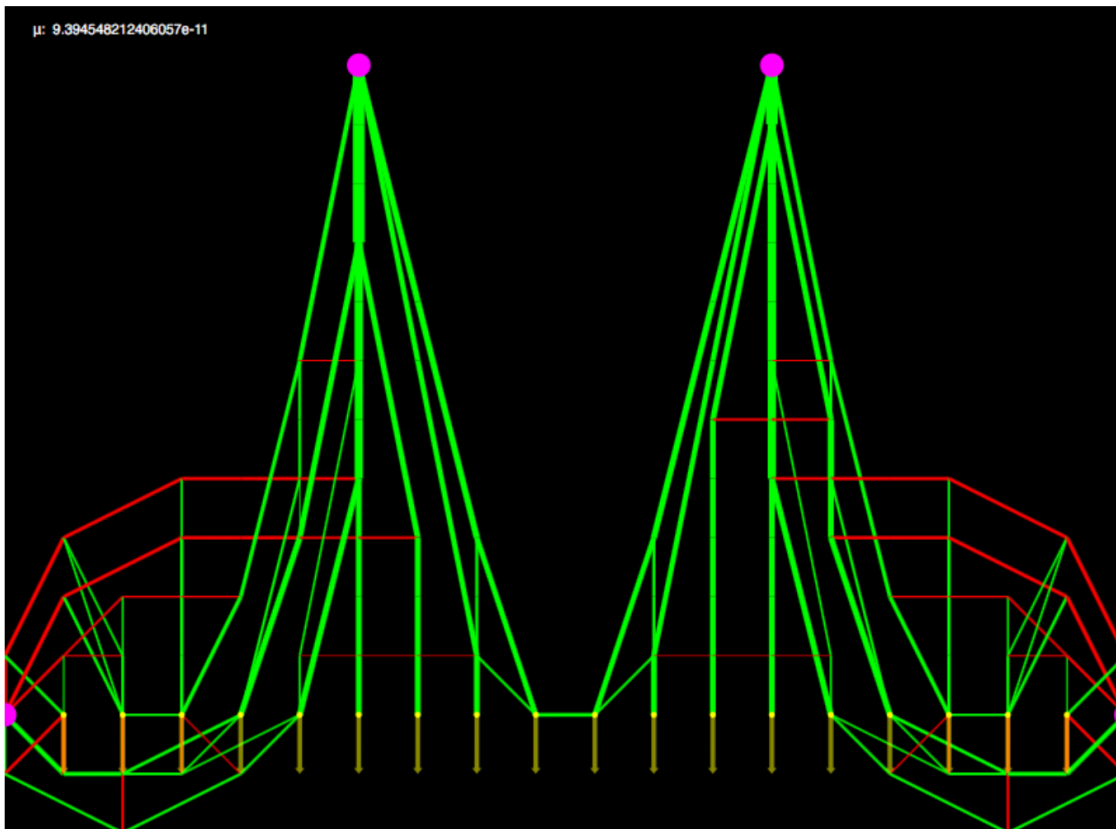
## via Structural Optimization

(Click [here](#) for older Java version.)

Height:  Width:

Use the mouse with x-load:  y-load:  ... or ... Select from this list:

Beam Strength:  Show:



# Algorithm Animation: Affine Scaling

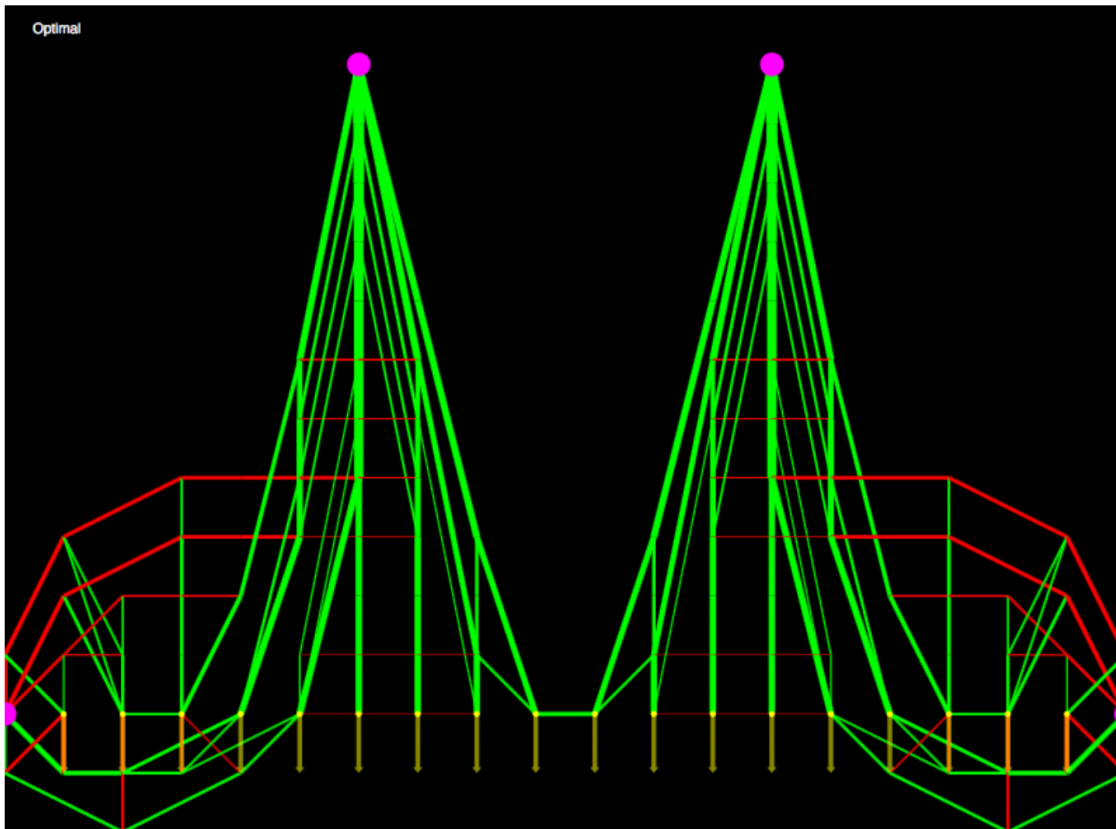
## via Structural Optimization

(Click [here](#) for older Java version.)

Height:  Width:  Neighbors:

Use the mouse with x-load:  y-load:  ... or ... Select from list

Step Length:  Beam Strength:



# Observations

- Basic optimal solutions lack geometric symmetry.
- The Two-Phase Simplex Method finds a feasible solution in about half of solution time.
- The feasible-but-not-optimal solutions look terrible (and bizarre).
- The Affine Scaling algorithm looks good all along but is not technically feasible until the very end.