

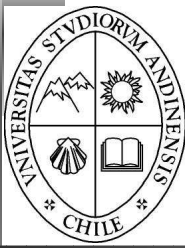
Optimization of trusses under uncertain loads.

Felipe Alvarez*, Miguel Carrasco[†], Benjamin Ivorra^{††}

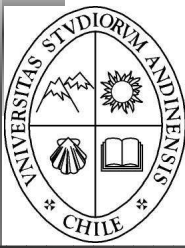
*Universidad de Chile, Departamento de Ingeniería Matemática

[†]Universidad de los Andes, Departamento de Ingeniería

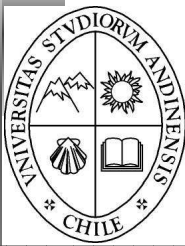
^{††}Universidad de Complutense de Madrid, Departamento de Matemáticas Aplicadas.



- **Optimization of trusses**
 - Standard minimum compliance truss design.
 - Instabilities and the standard multiload model.
 - **Random Perturbations: Minimizing the expected compliance.**
 - Numerical examples.
 - **Random Perturbations: Minimizing the variance.**
 - Numerical examples.

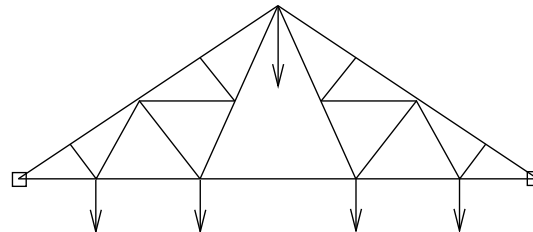


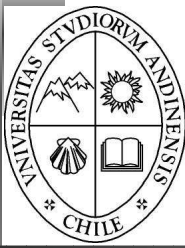
Optimization of trusses



Introduction

- Definition.
- Problem: find the **best structure** able to carry a external nodal force.
- Constraints: mechanical equilibrium , total volume and others . . .
- Minimize the compliance.



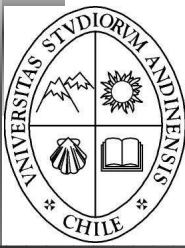


Minimum compliance truss design

(D)

$$\begin{aligned} \min_{\lambda, u} \quad & \frac{1}{2} f^T u \\ \text{s.t.} \quad & K(\lambda)u = f \\ & \lambda \in \Delta_m \end{aligned}$$

$\frac{1}{2} f^T u$ is called *Compliance*.

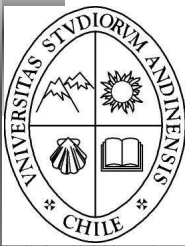


Mathematical formulation

- $\lambda \in \mathbb{R}^m$; m number of bars.
- n grades of freedom.
- $f \in \mathbb{R}^n$ nodal load vector.
- $u \in \mathbb{R}^n$ nodal displacements.

$$\begin{aligned} \min_{\lambda, u} \quad & \frac{1}{2} f^T u \\ \text{s.t.} \quad & K(\lambda)u = f \\ & \lambda \in \Delta_m \end{aligned}$$

- $\Delta_m = \{\lambda \in \mathbb{R}^m \mid \lambda \geq 0, \sum_{i=1}^m \lambda_i = 1\}.$
- $K(\lambda) \in \mathbb{R}^{n \times n}$, *stiffness matrix*.



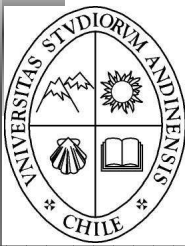
Stiffness Matrix

$$K(\lambda) = \sum_{i=1}^m \lambda_i K_i$$

where

$$K_i = b_i b_i^T \in \mathbb{R}^{n \times n}$$

K_i is dyadic.



Stiffness Matrix

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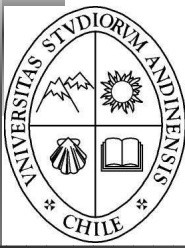
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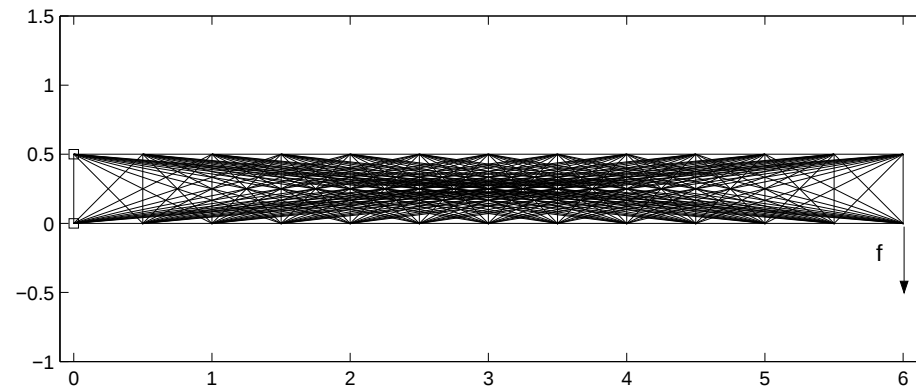
K_i is dyadic.

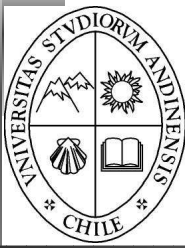
$$b_i = \frac{\sqrt{E_i}}{l_i} \gamma_i \in \mathbb{R}^n$$

- E_i : Young modulus.
- l_i : length of bar i .
- γ_i : *cosines/sines* vector.

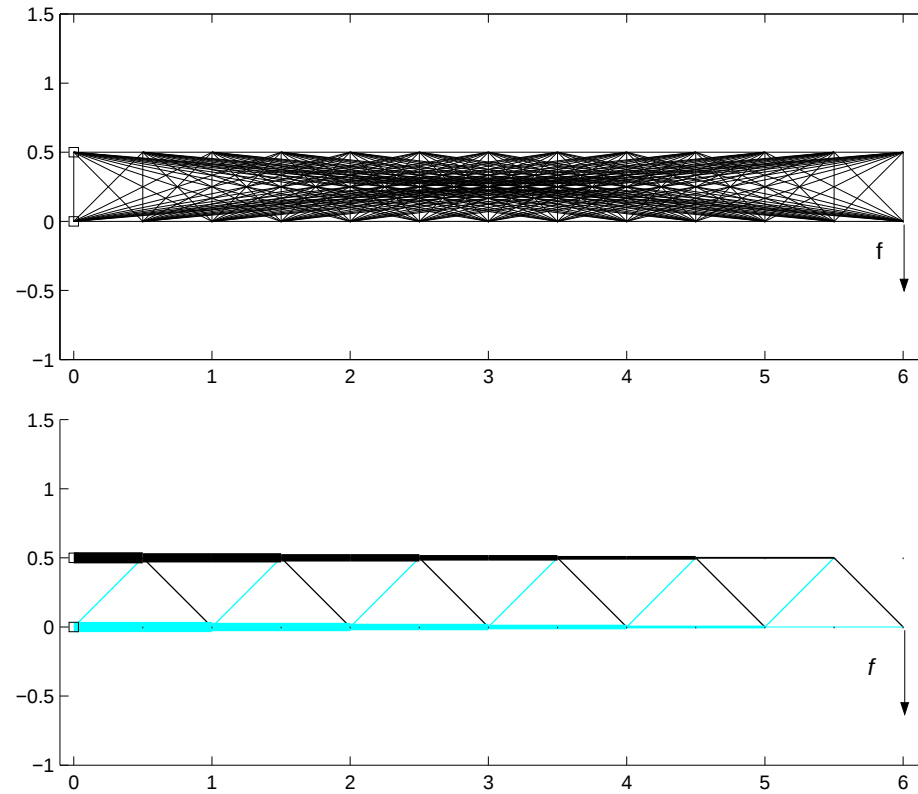


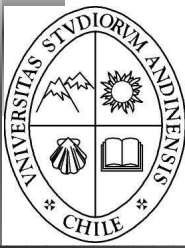
Ground structure approach



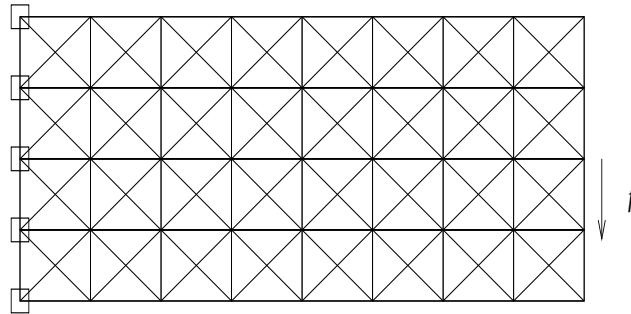


Ground structure approach

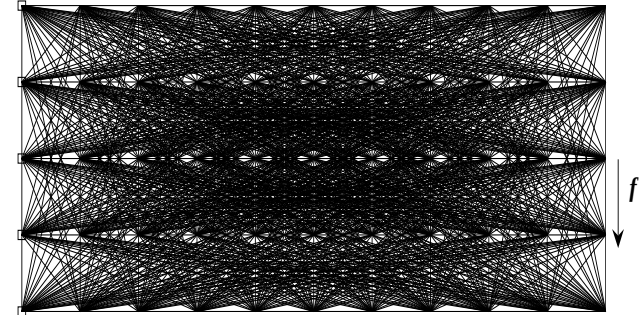
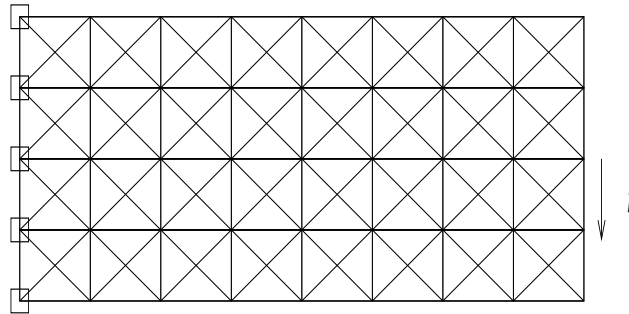




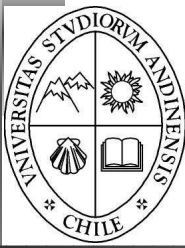
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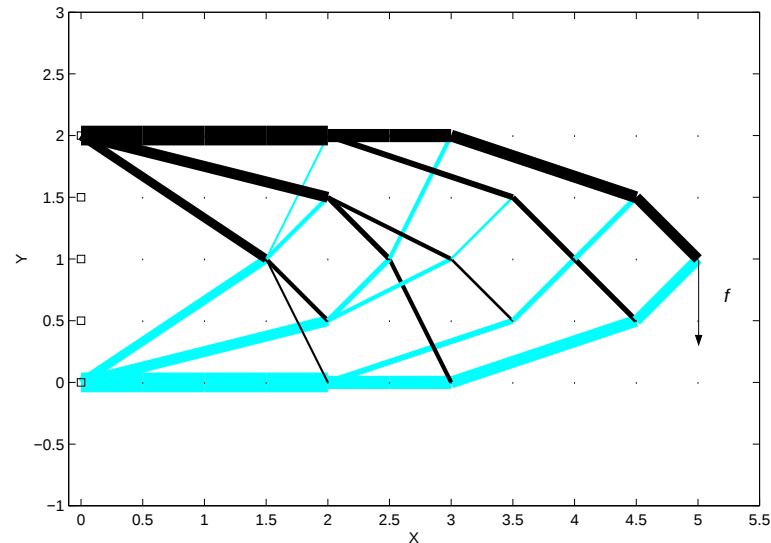
Ground structure approach



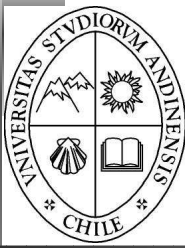
- Nodal positions are fixed in the reference configuration.
- We use a mesh full of nodes and bars.



Ground structure approach



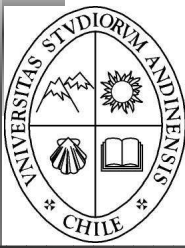
Typically a large number of bar vanish at the optimum.



(D)

$$\begin{aligned} \min_{\lambda \in \mathbb{R}^m, u \in \mathbb{R}^n} \quad & \frac{1}{2} f^T u \\ \text{s.t.} \quad & K(\lambda)u = f \\ & \lambda \in \Delta_m \end{aligned}$$

$\frac{1}{2} f^T u$ is independent of u satisfying $K(\lambda)u = f$.



Minimax Formulation

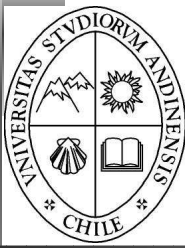
$$(\mathcal{D}) \quad \min_{\lambda \in \Delta_m} \left\{ \frac{1}{2} f^T u \mid K(\lambda)u = f \right\}$$

$$(\mathcal{D}) \quad \min_{\lambda \in \Delta_m} \max_{x \in \mathbb{R}^n} \left\{ f^T x - \frac{1}{2} x^T K(\lambda) x \right\}$$

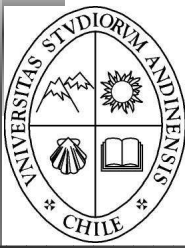


(Minimax Theorem)

$$(\mathcal{P}) \quad \min_{x \in \mathbb{R}^n} \max_{1 \leq i \leq m} \left\{ \frac{1}{2} x^T K_i x - f^T x \right\}$$

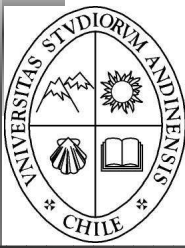


Instabilities and the standard multiload model



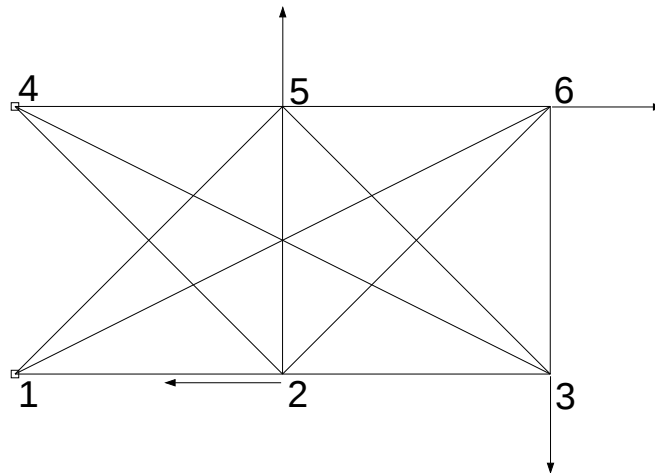
Example

The design problem ($\mathcal{D}$) may produce unsatisfactory results in respect to mechanical stability (see Ben-Tal and Nemirovsky 1997)

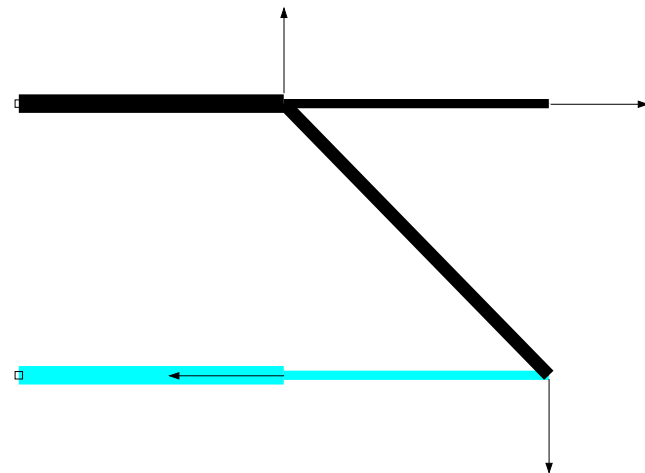
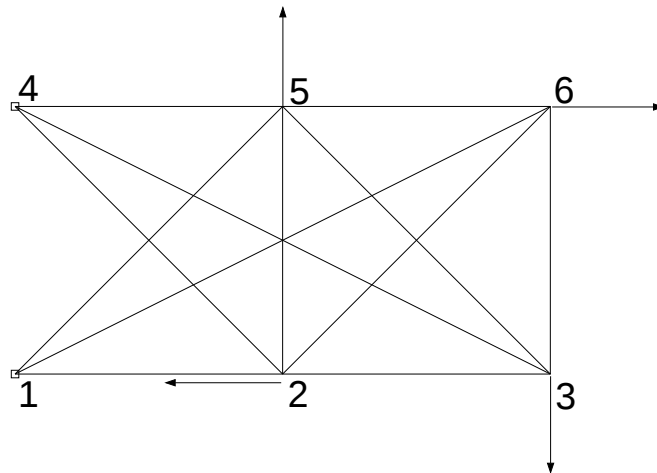


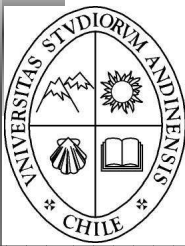
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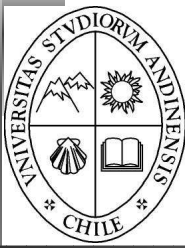
Standard multiload model

$$\min_{\lambda \in \mathbb{R}^m} \frac{1}{2} \sum_{j=1}^k \alpha_j f^{jT} u^j$$

$$s.t. \quad K(\lambda) u^j = f^j, \quad j = 1, \dots, k$$

$$\lambda \in \Delta_m$$

We minimize a weighted average of the compliances.



Defining

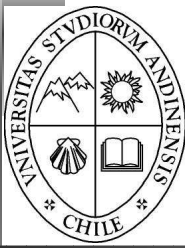
$$\hat{f} = (\alpha_1 f^{1T}, \dots, \alpha_k f^{kT})^T \in \mathbb{R}^{n(k+1)},$$

$$\hat{K}_i = \text{diag}(\alpha_1 K_i, \dots, \alpha_k K_i) \in \mathbb{R}^{n(k+1) \times n(k+1)},$$

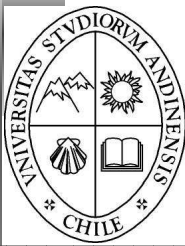
$$\hat{K}(\lambda) = \sum_{i=1}^m \lambda_i \hat{K}_i,$$

Then multiload model can be written as $(\mathcal{D})$.

remark: $\hat{K} \neq b_i b_i^T$.



Random Perturbations: Minimizing the expected compliance.



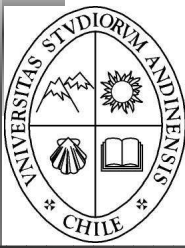
Random load model

Let $\xi \in \mathbb{R}^n$ be a perturbation on the load vector f .

$$\Psi(\xi, \lambda) = \begin{cases} \frac{1}{2}(f + \xi)^T x & \text{if } \lambda \in \Delta_m \text{ and } u \in \mathbb{R}^n \text{ such that} \\ & K(\lambda)u = f + \xi, \\ +\infty & \text{otherwise} \end{cases}$$

$$\Psi: (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n)) \rightarrow (\mathbb{R} \cup \{+\infty\}, \bar{\mathcal{B}}(\mathbb{R}))$$

Results to be proper, lower semicontinuous and convex.

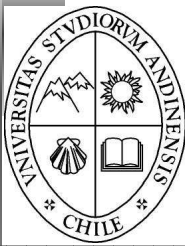


Random load model

For each $\lambda \in \Delta_m$

$$\begin{array}{ccc} \Psi(\cdot, \lambda): & (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n)) & \rightarrow (\mathbb{R} \cup \{+\infty\}, \bar{\mathcal{B}}(\mathbb{R})) \\ & \xi & \mapsto \Psi(\xi, \lambda) \end{array}$$

is measurable.



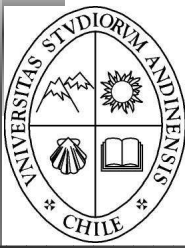
Random load model

Minimum expected compliance design problem

$$(\mathcal{D}; \mathbb{P}) \quad \min_{\lambda \in \Delta^m} \mathbb{E}_{\xi}[\Psi(\xi, \lambda)]$$

We assume that ξ is a random variable corresponding to an uncertain nodal load perturbation

$$\begin{aligned} \xi: \quad (\Omega, \mathcal{A}) &\longrightarrow (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n)) \\ \omega &\longmapsto \xi(\omega) \end{aligned}$$

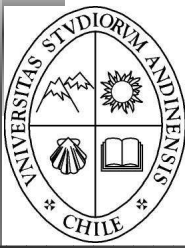


Example: discrete perturbation

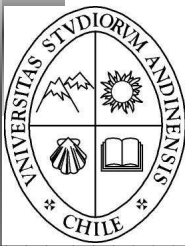
1. $\mathbb{P}$ with finite support $\text{Sop}(\mathbb{P}) = \{\xi^1, \dots, \xi^k\}$.

$$\begin{aligned} (\mathcal{D}; \mathbb{P}) \quad & \min_{\lambda \in \mathbb{R}^m} \frac{1}{2} \sum_{j=1}^k \alpha_j f^{jT} u^j \\ & s.t. \quad K(\lambda) u^j = f^j, \quad i = 1, \dots, k \\ & \quad \lambda \in \Delta_m \end{aligned}$$

$(\mathcal{D}; \mathbb{P})$ is the *standard multiload model* where $\alpha_j = \mathbb{P}(\xi = \xi^j)$ and $f^j = f + \xi^j$, $j = 1, \dots, k$.



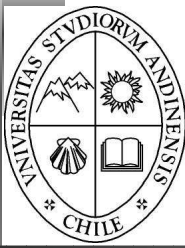
Continuous case



Continuous case

Theorem 0.1. Let $\xi: \Omega \rightarrow \mathbb{R}^n$ be a continuous random variable with mean vector $\mathbb{E}(\xi) = 0$ and Variance–covariance matrix $\text{Var}(\xi) = PP^T$, with $P \in \mathbb{R}^{n \times k}$.
Then the corresponding minimum expected compliance design problem $(\mathcal{D}; \mathbb{P})$ is given by

$$\begin{aligned} (\mathcal{D}; \mathbb{P}) \quad & \min_{\lambda \in \mathbb{R}^m} \underbrace{\frac{1}{2} f^T u}_{\text{mean load}} + \underbrace{\frac{1}{2} \text{Trace}(P^T U)}_{\text{variance}} \\ & s.t. \quad K(\lambda)u = f \\ & \quad \quad K(\lambda)U = P \\ & \quad \quad \lambda \in \Delta_m \end{aligned}$$



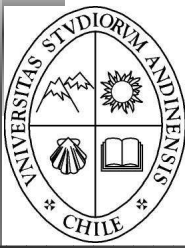
Remark p^j , $j = 1, \dots, k$, columns of P
Defining

$$\hat{f} = (f^T, p^{1T}, \dots, p^{kT})^T \in \mathbb{R}^{n(k+1)},$$

$$\hat{K}_i = \text{diag}(K_i, K_i, \dots, K_i) \in \mathbb{R}^{n(k+1) \times n(k+1)},$$

$$\hat{K}(\lambda) = \sum_{i=1}^m \lambda_i \hat{K}_i,$$

Then $(\mathcal{D}; \mathbb{P})$ may be written as a multiload model.



Random multidimensional independent perturbation

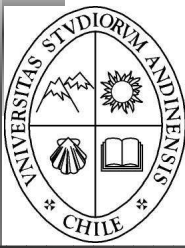
$\xi = \sum_{j=1}^k \varepsilon_j d^j$, where
 $d^j \in \mathbb{R}^n$, $\mathbb{E}(\varepsilon_j) = 0$ and $\text{Var}(\varepsilon_j) = \sigma_j^2$.

$$(\mathcal{D}; \mathbb{P}) \quad \min_{\lambda \in \mathbb{R}^m} \frac{1}{2} f^T u + \frac{1}{2} \sum_{j=1}^r \sigma_j d^j T u^j$$

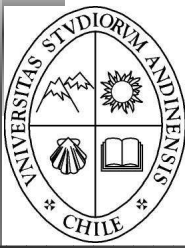
$$s.t. \quad K(\lambda)u = f$$

$$K(\lambda)u^j = \sigma_j d^j \quad j = 1, \dots, k$$

$$\lambda \in \Delta_m.$$

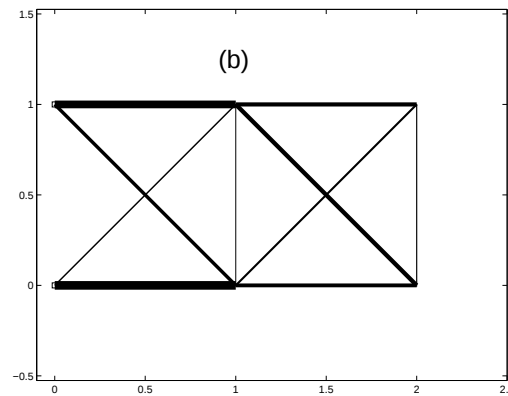
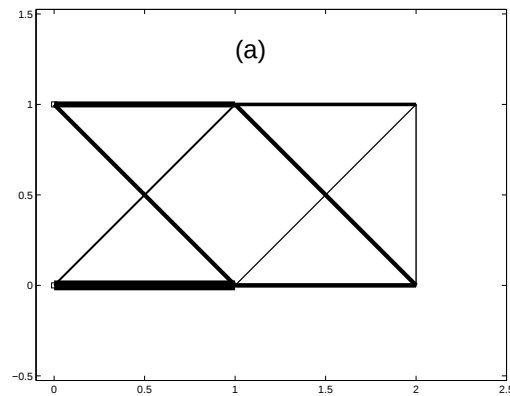
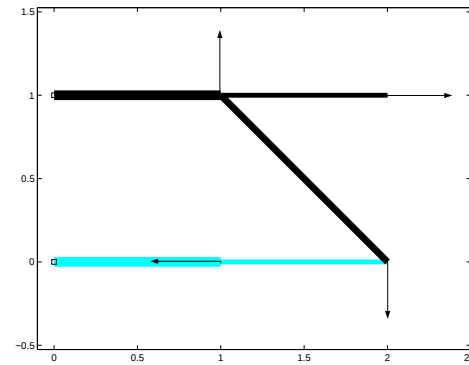
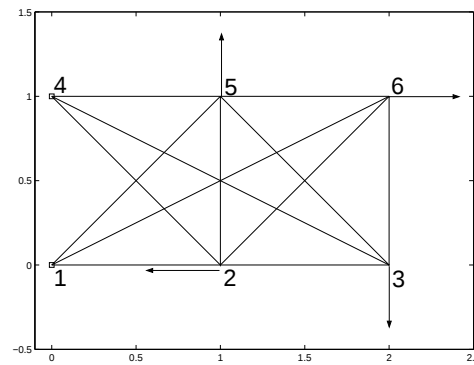


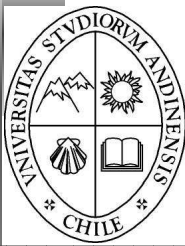
Numerical results



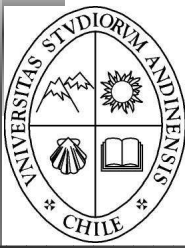
Toy example

Toy example, Ben-Tal and Nemirowski 1997

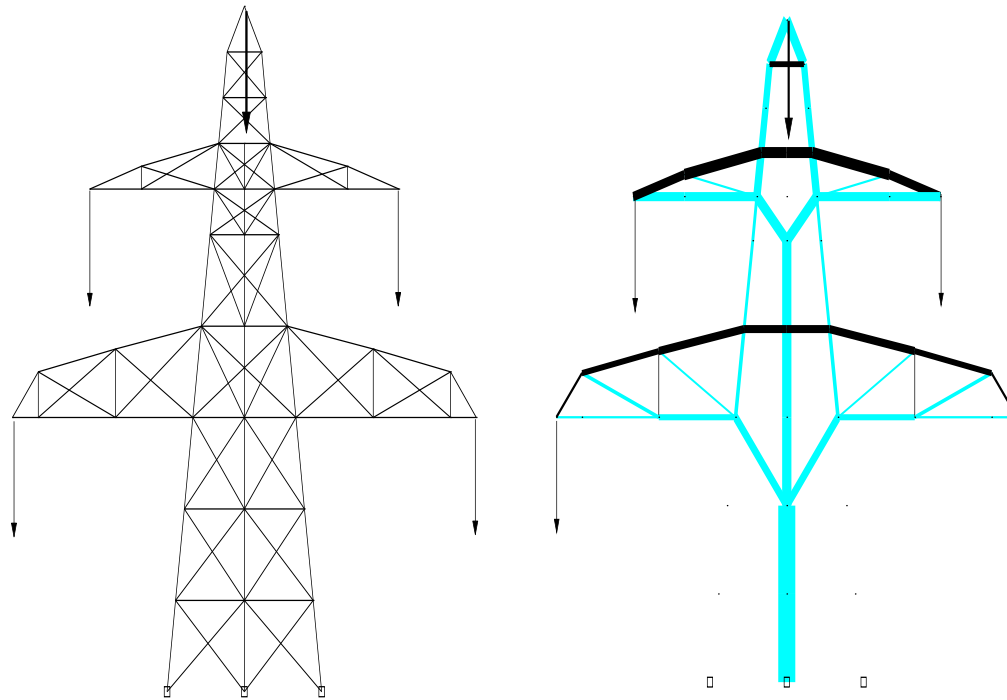


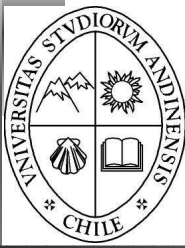


Problem	compl. mean	max compl.	min compl.
Standard single load	$\nexists$	$+\infty$	$+\infty$
(a)	0.025	0.053	0.013
(b)	0.023	0.051	0.012

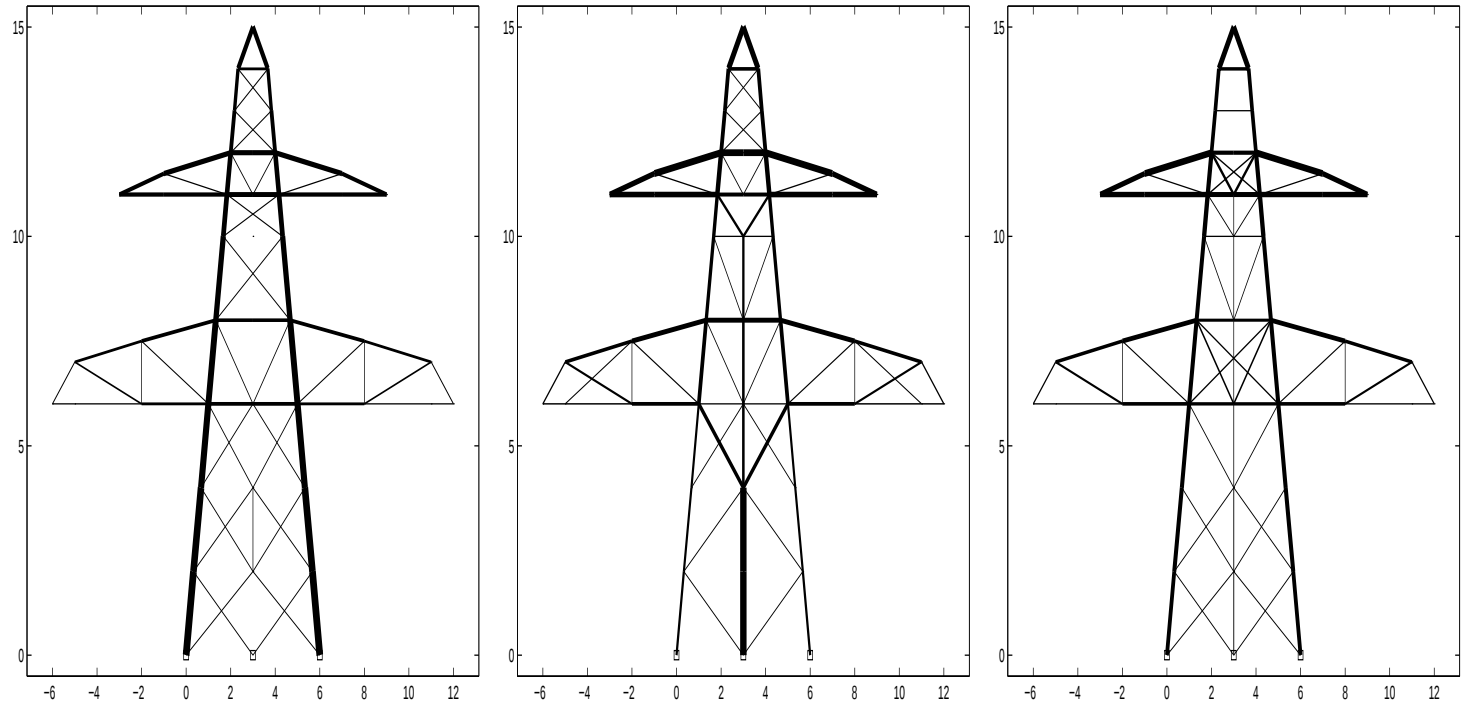


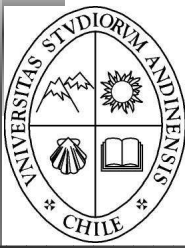
Electricity mast, Achtziger et al. 1992



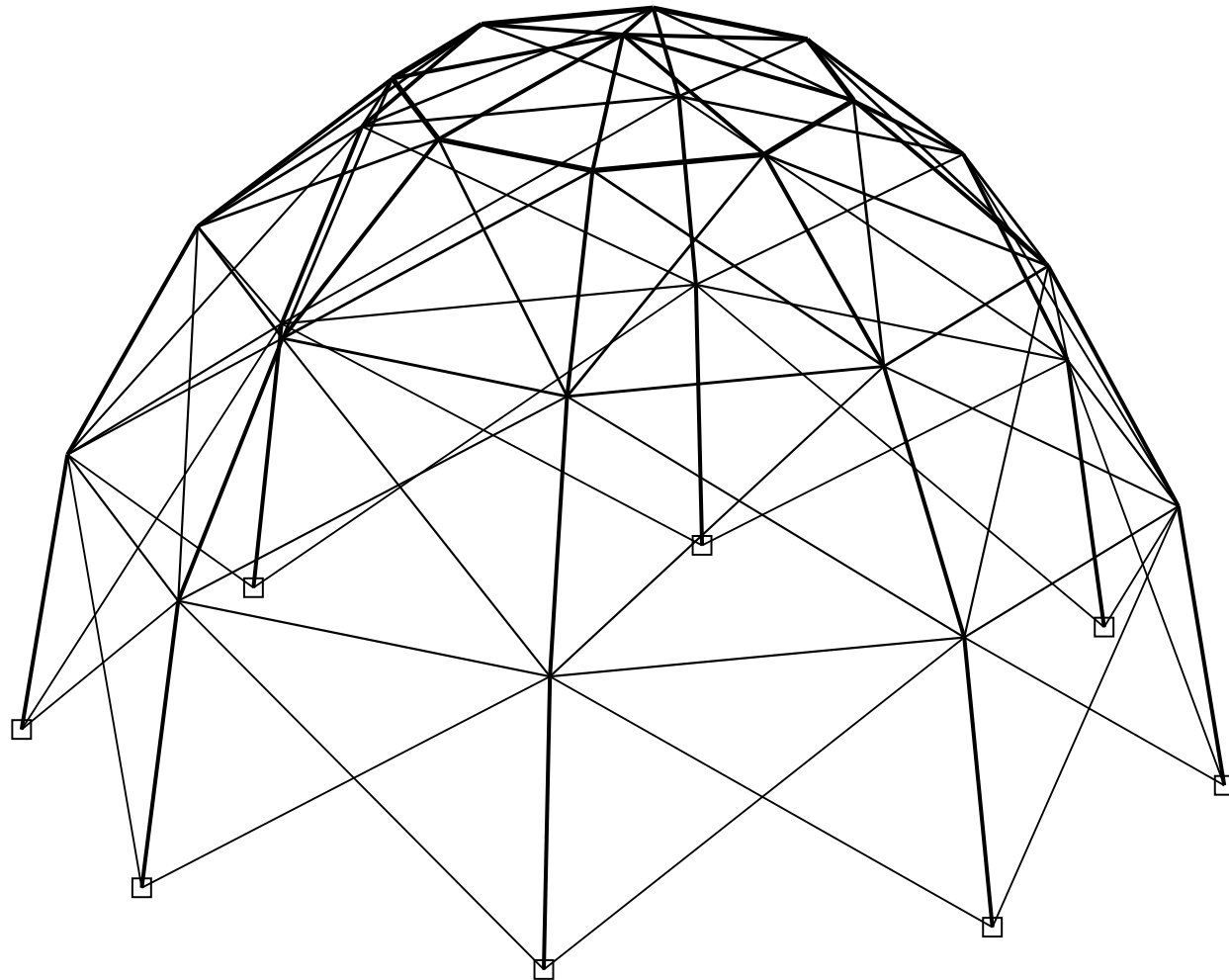


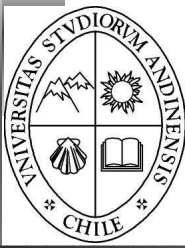
Electricity mast





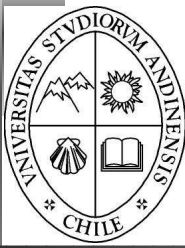
Dome





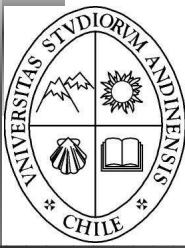
Dome



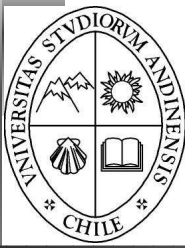


Dome





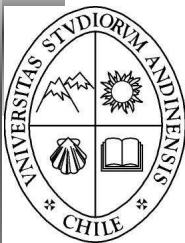
Random Perturbations: Minimizing the variance.



Stochastic model including variance

We consider

$$\min_{\lambda \in \mathbb{R}^m} \alpha \mathbb{E}_{\xi}[\Psi(\xi, \lambda)] + \beta \text{Var}[\Psi(\xi, \lambda)].$$



Stochastic model including variance

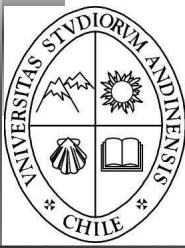
We consider

$$\min_{\lambda \in \mathbb{R}^m} \alpha \mathbb{E}_{\xi}[\Psi(\xi, \lambda)] + \beta \text{Var}[\Psi(\xi, \lambda)]$$

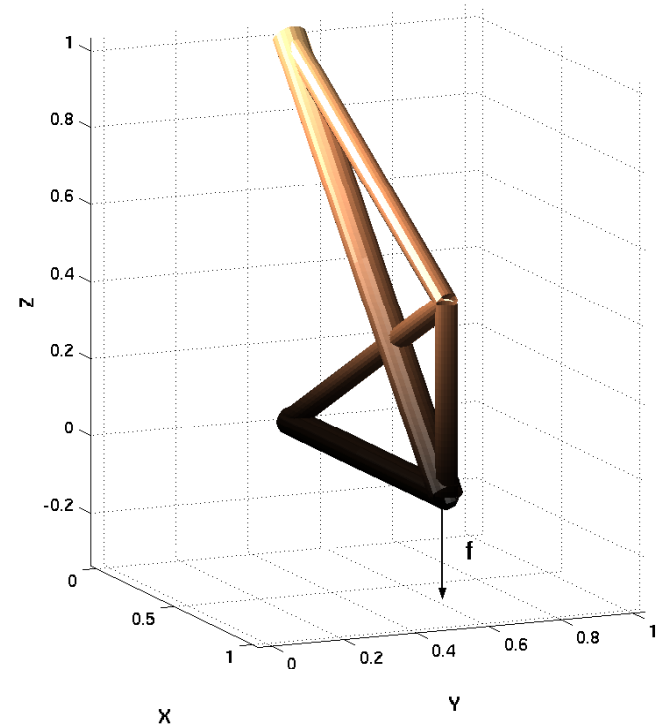
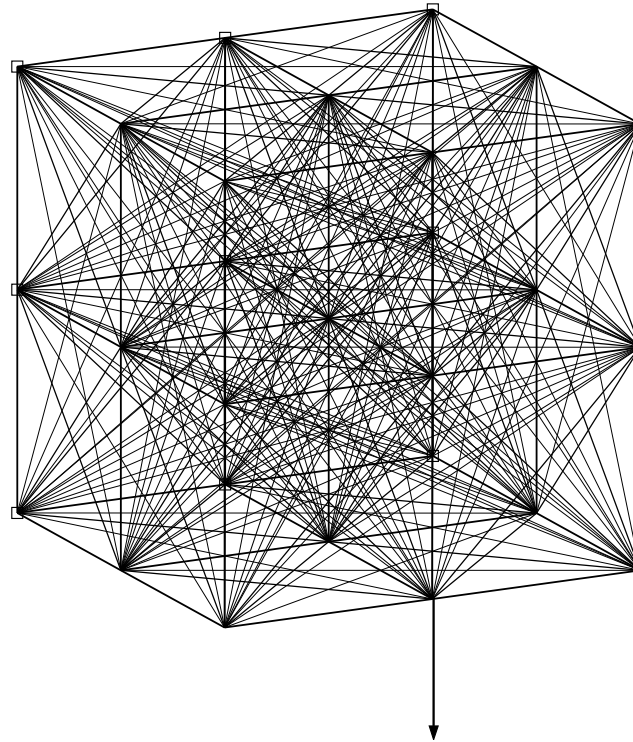
Theorem 0.2. $\xi \sim \mathcal{N}(0, PP^T)$. Taking for simplicity $\alpha = 0$ and $\beta = 1$.

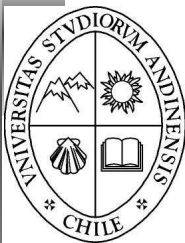
$$\begin{aligned} (\mathcal{D}; \mathbb{P}_{var}) \quad & \min_{\lambda \in \mathbb{R}^m} \frac{1}{2} \left(\text{Trace}((P^T U)^2) + 2f^T U U^T f \right) \\ & s.t. \quad K(\lambda)u = f \\ & \quad \quad K(\lambda)U = P \\ & \quad \quad \lambda \in \Delta_m \end{aligned}$$

Apparent harder to solve.

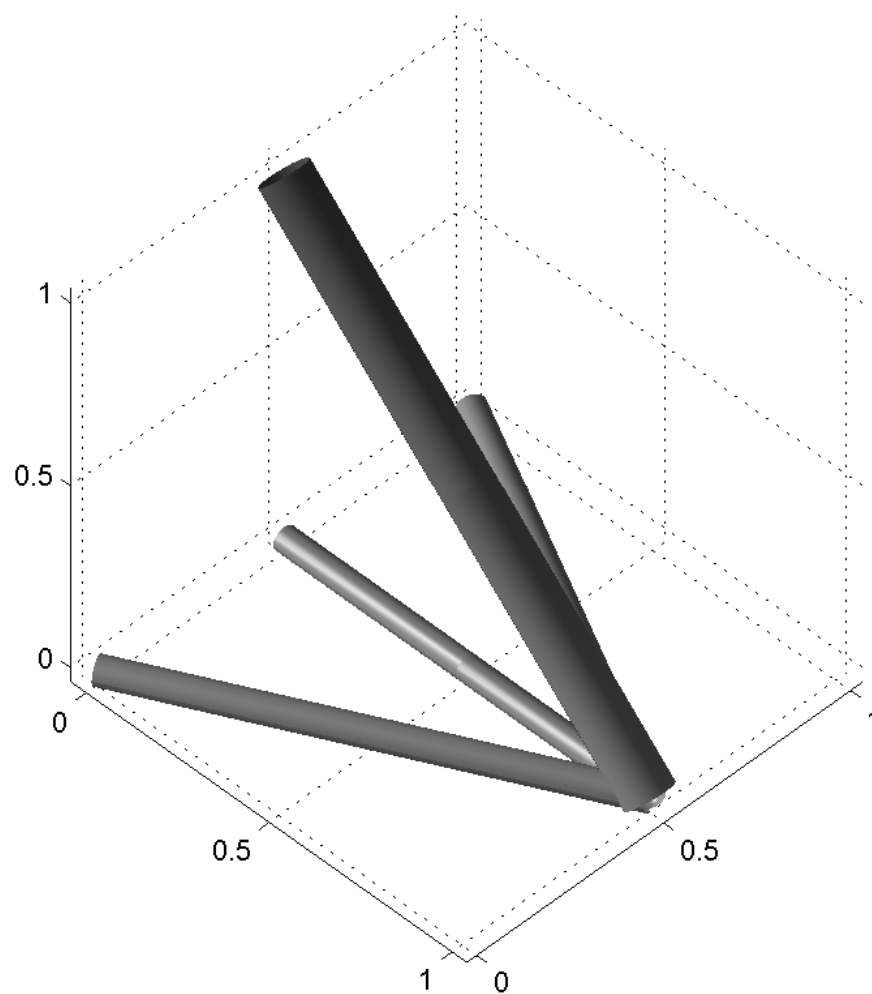


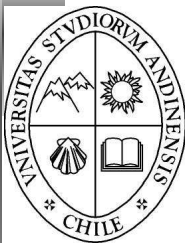
Michel 3D



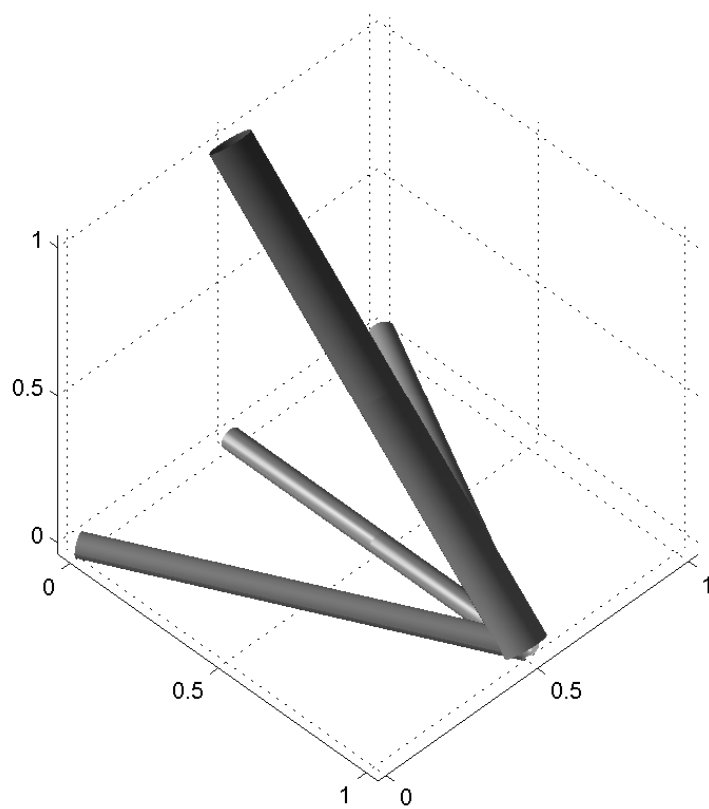


$\alpha=1$ and $\beta=0$

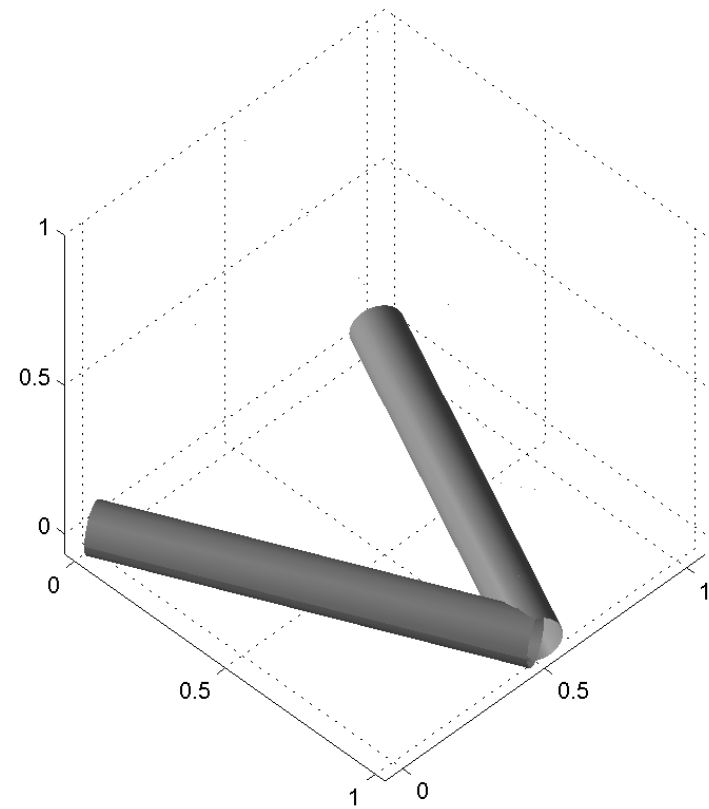




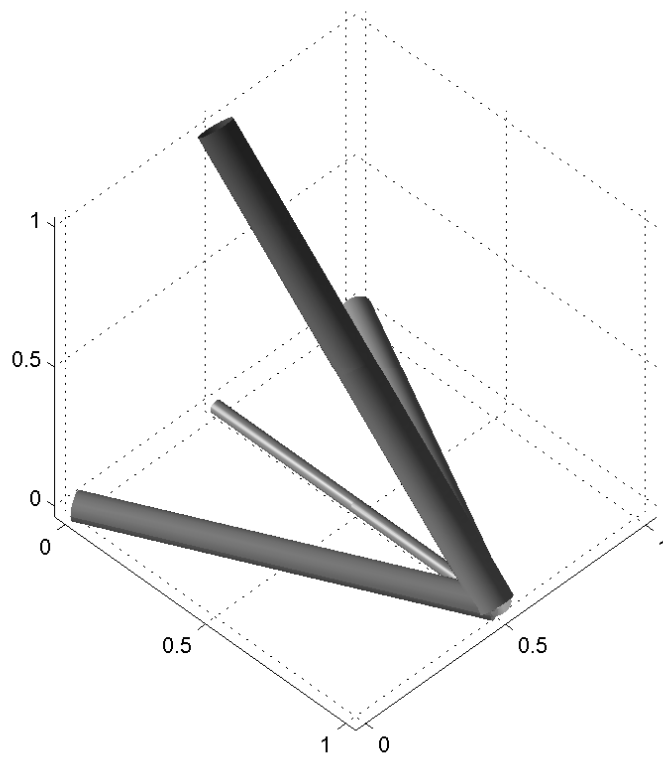
$\alpha=1$ and $\beta=0$



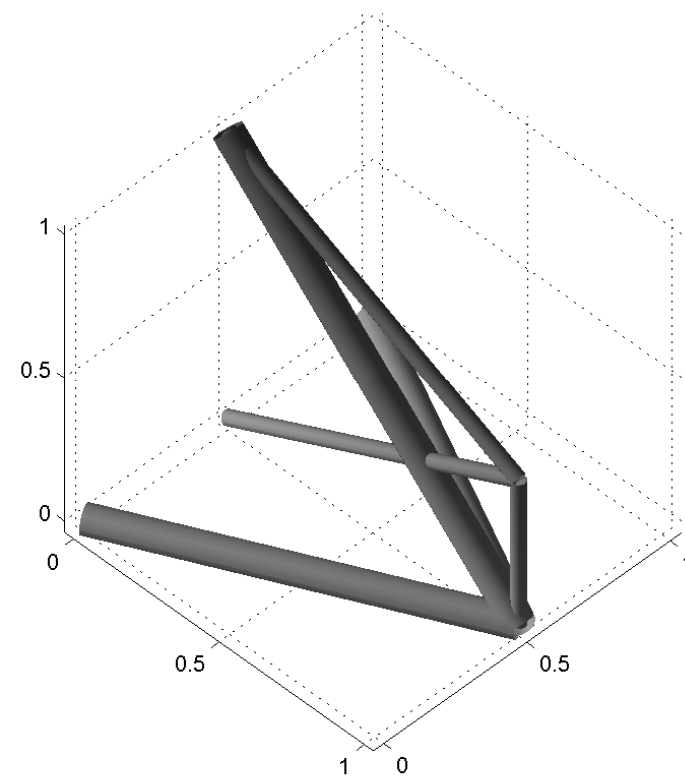
$\alpha=0$ and $\beta=1$

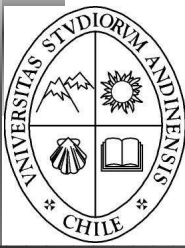


$\alpha=1$ and $\beta=0.25$

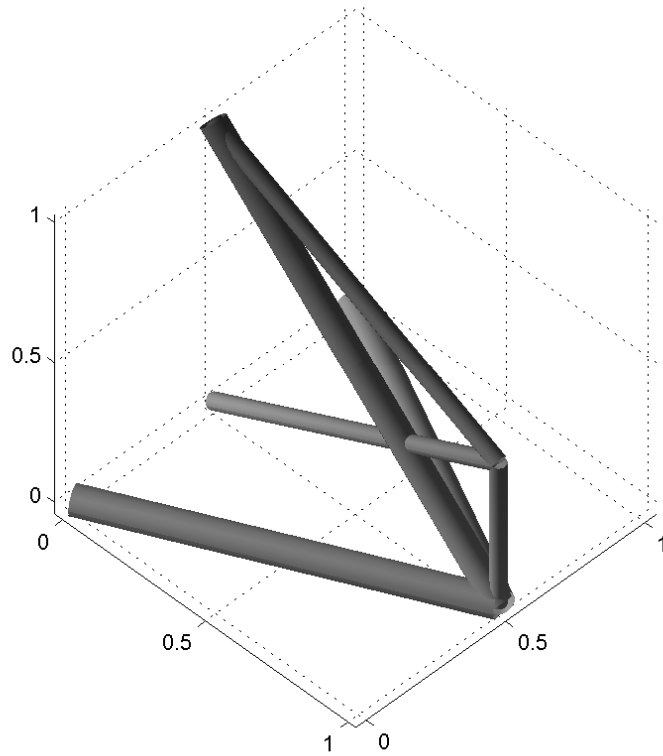


$\alpha=1$ and $\beta=0.5$

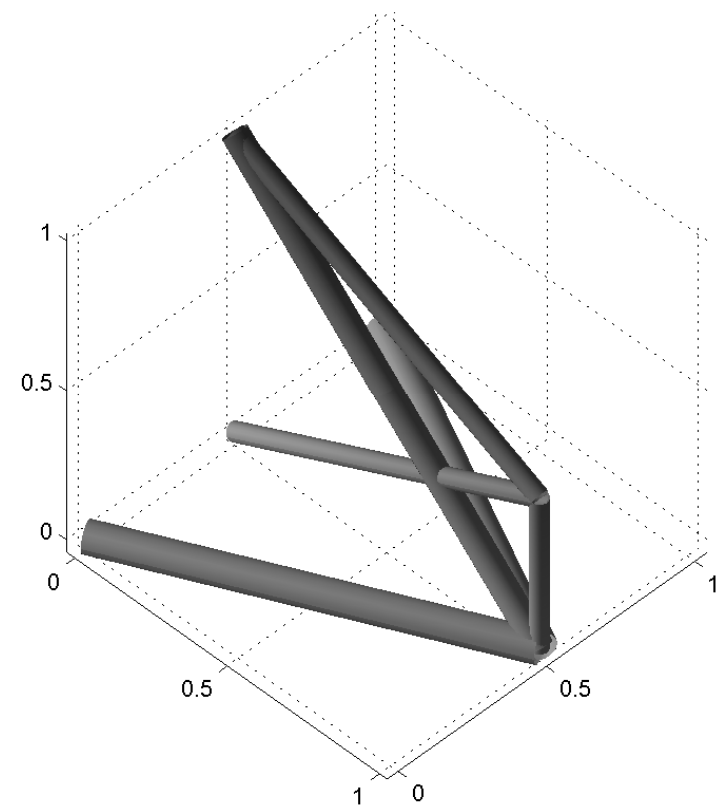


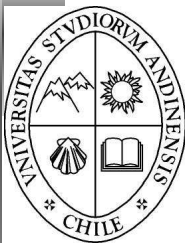


$\alpha=1$ and $\beta=0.75$

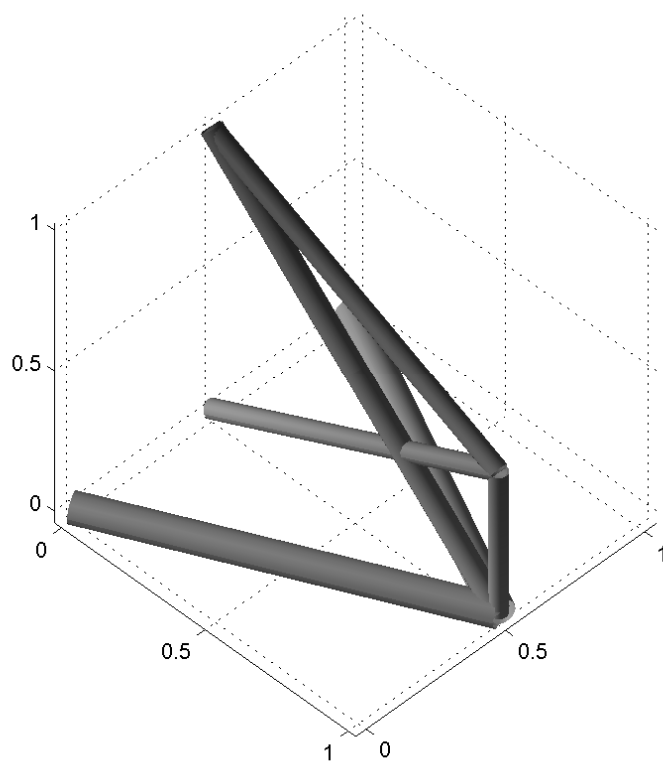


$\alpha=1$ and $\beta=1$

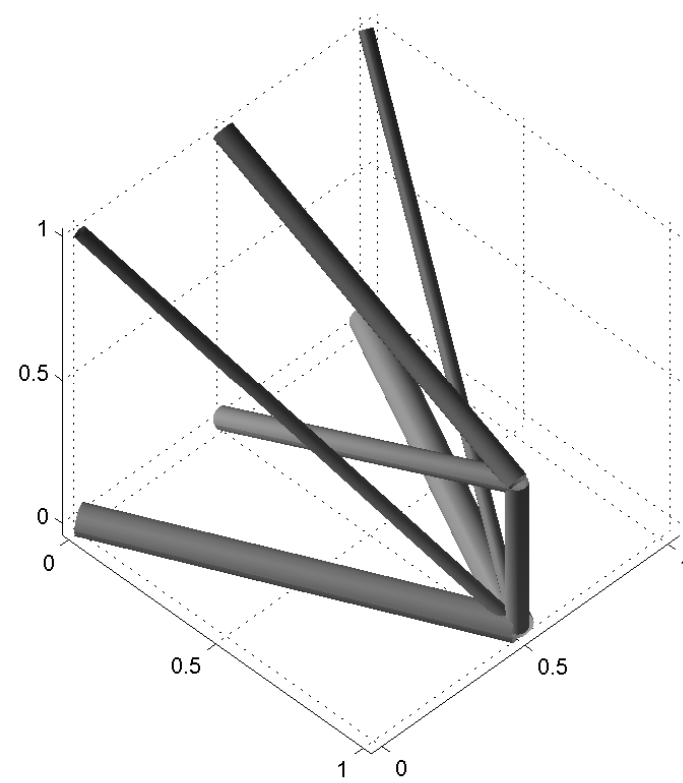


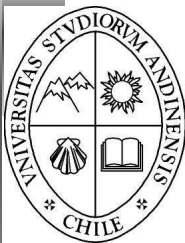


$\alpha=0.75$ and $\beta=1$

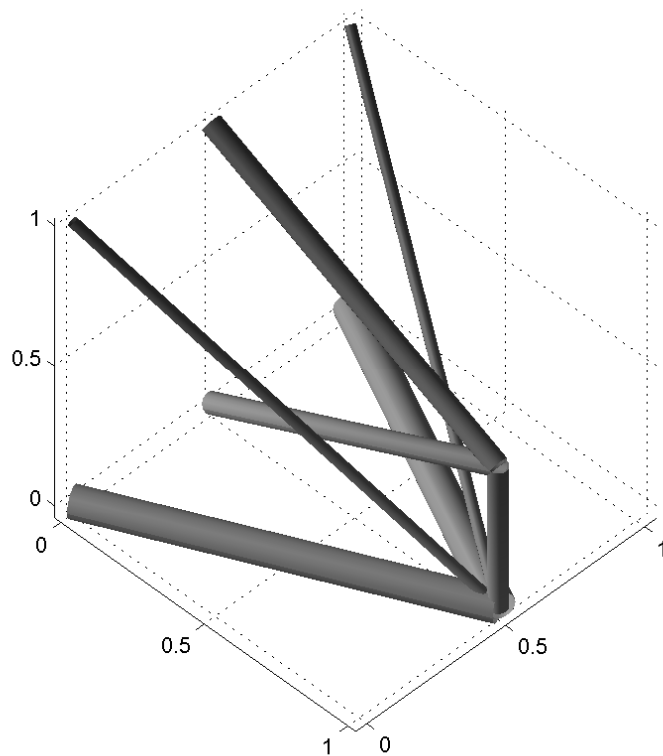


$\alpha=0.5$ and $\beta=1$

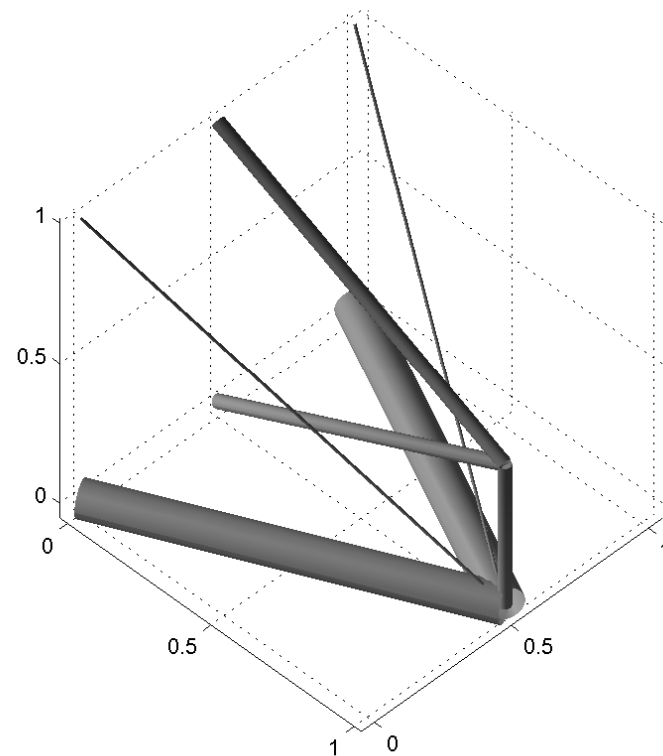


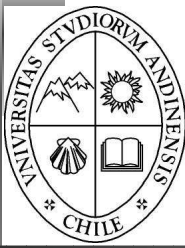


$\alpha=0.25$ and $\beta=1$

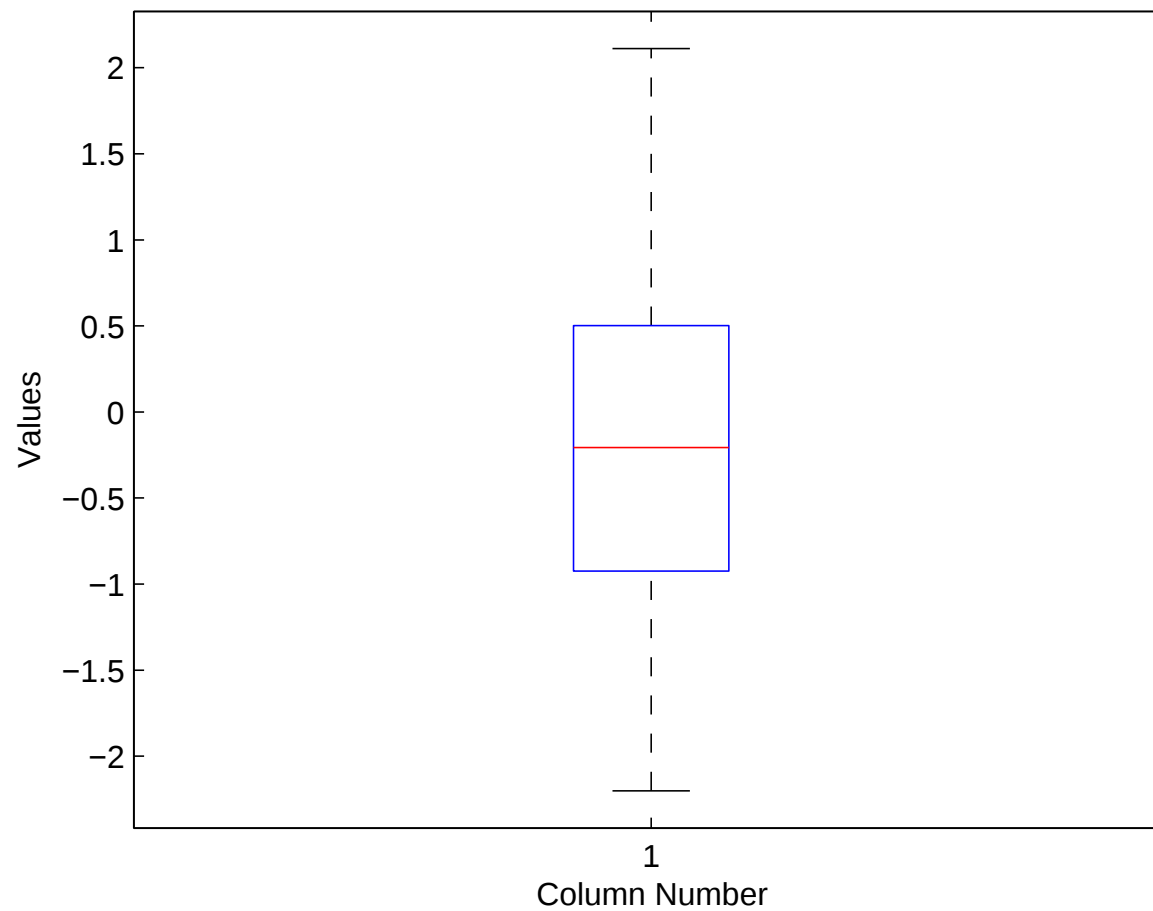


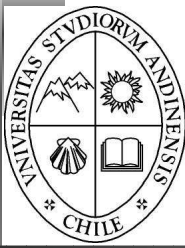
$\alpha=0.01$ and $\beta=1$



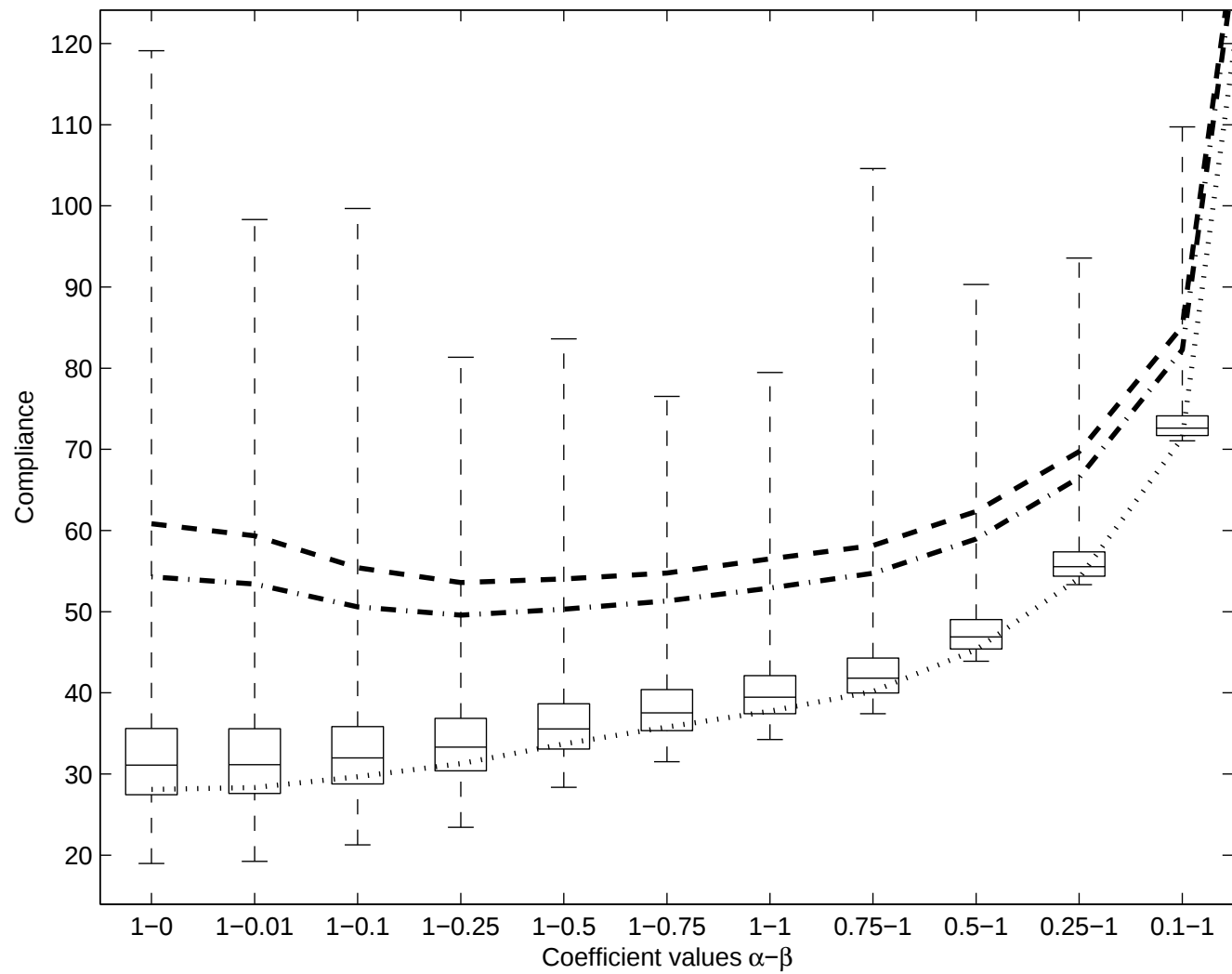


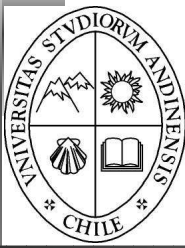
Boxplot





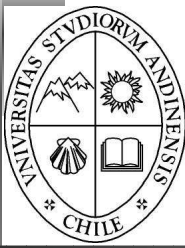
Boxplot





Conclusions

- Classical model (single load) don't get good results .
- Random model helps to avoid this problem.
- It's equivalent to multiload one, but another interpretations in weight factors holds.
- Continuous Model.
- We have to know the influence of each perturbation.
- Increase the dimension of the problem.
- Highly non linear.



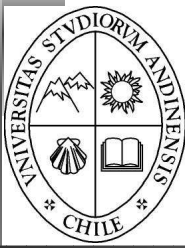
Optimization of trusses under uncertain loads.

Felipe Alvarez*, Miguel Carrasco[†], Benjamin Ivorra^{††}

*Universidad de Chile, Departamento de Ingeniería Matemática

[†]Universidad de los Andes, Departamento de Ingeniería

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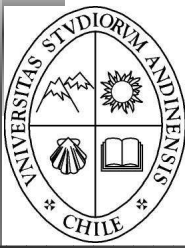
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Boxplot

