

# The Euclidean Steiner Tree Problem in $\mathbb{R}^n$

## *Mathematical Models*

N. Maculan<sup>‡</sup>, V. Costa<sup>§</sup>

Universidade Federal do Rio de Janeiro  
COPPE – Programa de Engenharia de Sistemas

---

<sup>‡</sup> maculan@cos.ufrj.br  
<sup>§</sup> virscosta@gmail.com

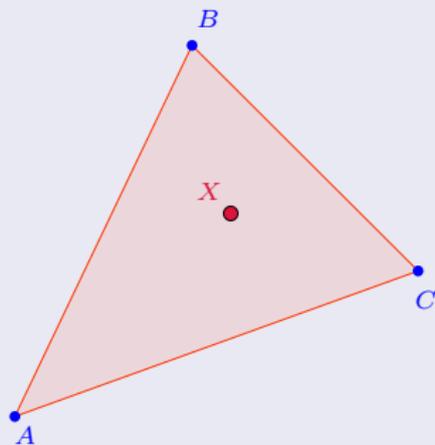
- 1 Problem Definition
- 2 Properties
- 3 First Formulation
- 4 Second Formulation
- 5 Second Formulation: Experiments on Platonic Solids

# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Triangle: Three given points

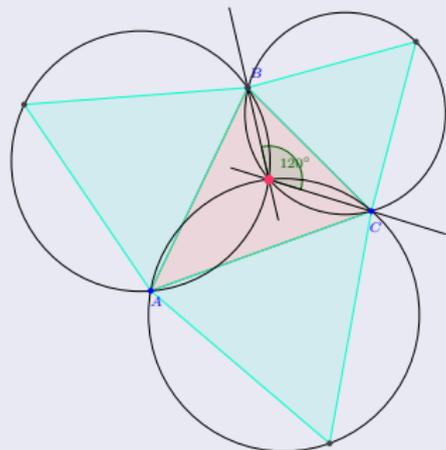


# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Triangle: Three given points

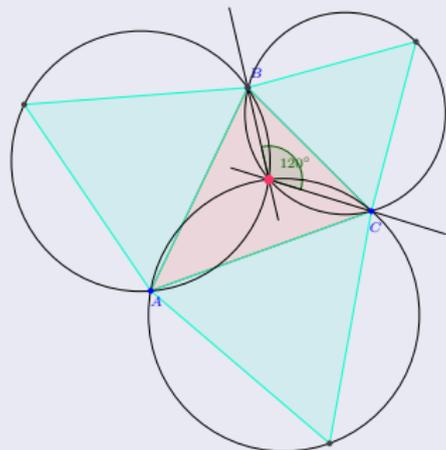


# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Triangle: Three given points

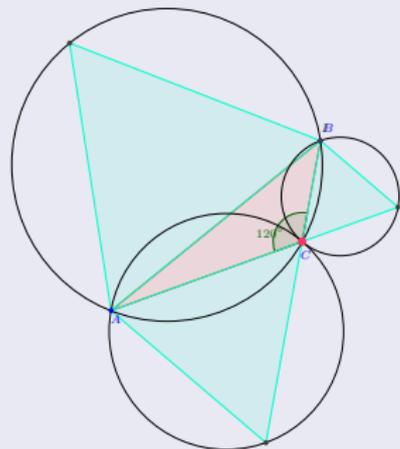


**Torricelli** (1647) pointed out a solution when the triangle formed by the three given points does not have an angle  $\geq 120^\circ$ .

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Triangle: Three given points



**Torricelli** (1647) pointed out a solution when the triangle formed by the three given points does not have an angle  $\geq 120^\circ$ .

**Heinen** (1837) apparently is the first to prove that, for a triangle in which an angle is  $\geq 120^\circ$ , the vertex associated with this angle is the minimizing point.



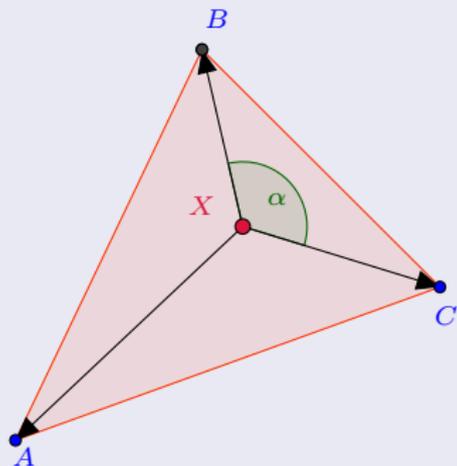
# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Fermat's Challenge as an Optimization Problem

$$\text{Minimize } \mathcal{D} = \|\vec{XA}\| + \|\vec{XB}\| + \|\vec{XC}\|$$

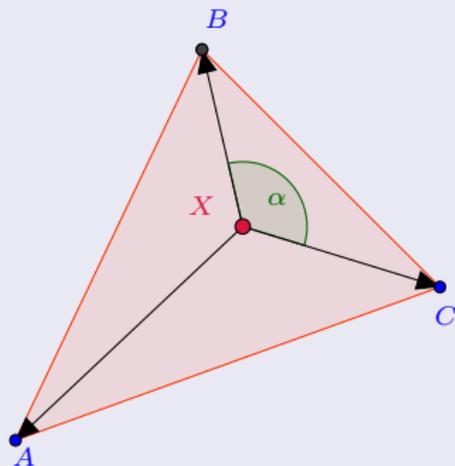


# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Fermat's Challenge as an Optimization Problem



$$\text{Minimize } \mathcal{D} = \|\vec{XA}\| + \|\vec{XB}\| + \|\vec{XC}\|$$

The solution is given when

$$\nabla \mathcal{D} = 0.$$

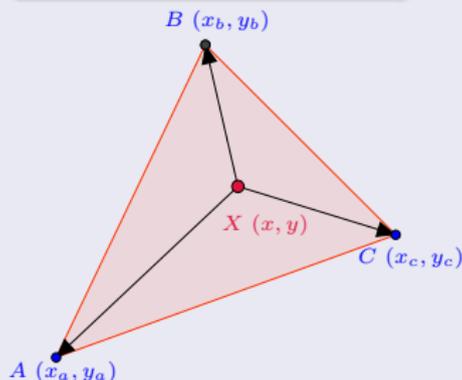
# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Fermat's Challenge as an Optimization Problem

$$\text{Min } \mathcal{D} = \|\vec{XA}\| + \|\vec{XB}\| + \|\vec{XC}\|$$



$$\|\vec{XA}\| = \sqrt{(x_a - x)^2 + (y_a - y)^2}$$

$$\|\vec{XB}\| = \sqrt{(x_b - x)^2 + (y_b - y)^2}$$

$$\|\vec{XC}\| = \sqrt{(x_c - x)^2 + (y_c - y)^2}$$

$$\nabla \mathcal{D} = \begin{pmatrix} \frac{\partial \mathcal{D}}{\partial x} \\ \frac{\partial \mathcal{D}}{\partial y} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

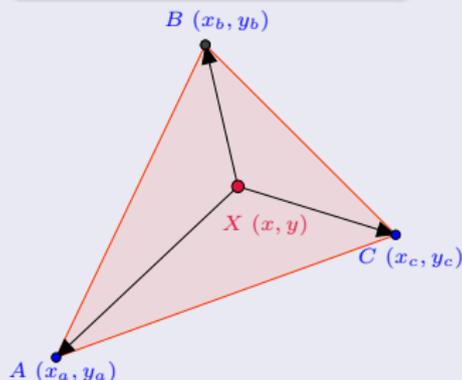


## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Fermat's Challenge as an Optimization Problem

$$\text{Min } \mathcal{D} = \|\vec{XA}\| + \|\vec{XB}\| + \|\vec{XC}\|$$



$$\frac{\partial \mathcal{D}}{\partial x} = \frac{x_a - x}{\|\vec{XA}\|} + \frac{x_b - x}{\|\vec{XB}\|} + \frac{x_c - x}{\|\vec{XC}\|} = 0$$

$$\frac{\partial \mathcal{D}}{\partial y} = \frac{y_a - y}{\|\vec{XA}\|} + \frac{y_b - y}{\|\vec{XB}\|} + \frac{y_c - y}{\|\vec{XC}\|} = 0$$

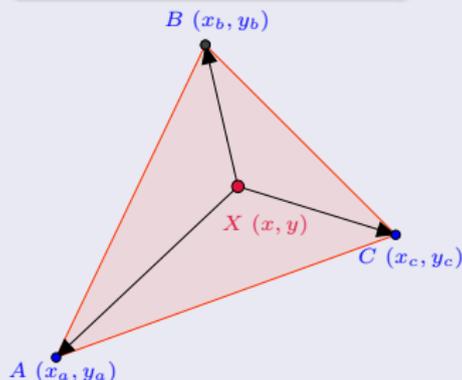
# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Fermat's Challenge as an Optimization Problem

$$\text{Min } \mathcal{D} = \|\vec{XA}\| + \|\vec{XB}\| + \|\vec{XC}\|$$



$$\begin{pmatrix} \frac{\partial \mathcal{D}}{\partial x} \\ \frac{\partial \mathcal{D}}{\partial y} \end{pmatrix} = \underbrace{\begin{pmatrix} \frac{x_a - x}{\|\vec{XA}\|} \\ \frac{y_a - y}{\|\vec{XA}\|} \end{pmatrix} + \begin{pmatrix} \frac{x_b - x}{\|\vec{XB}\|} \\ \frac{y_b - y}{\|\vec{XB}\|} \end{pmatrix} + \begin{pmatrix} \frac{x_c - x}{\|\vec{XC}\|} \\ \frac{y_c - y}{\|\vec{XC}\|} \end{pmatrix}}_{\text{Unitary Vectors Sum}} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

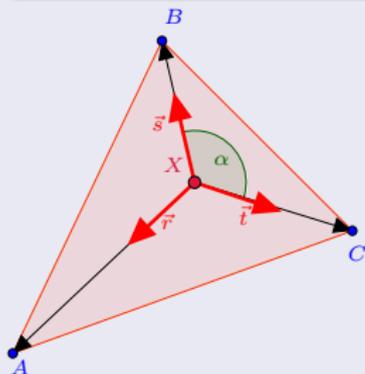
# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Fermat's Challenge as an Optimization Problem

$$\text{Min } \mathcal{D} = ||\vec{XA}|| + ||\vec{XB}|| + ||\vec{XC}||$$



Three Forces in Equilibrium

$$\nabla \mathcal{D} = \vec{r} + \vec{s} + \vec{t} = \vec{0}$$

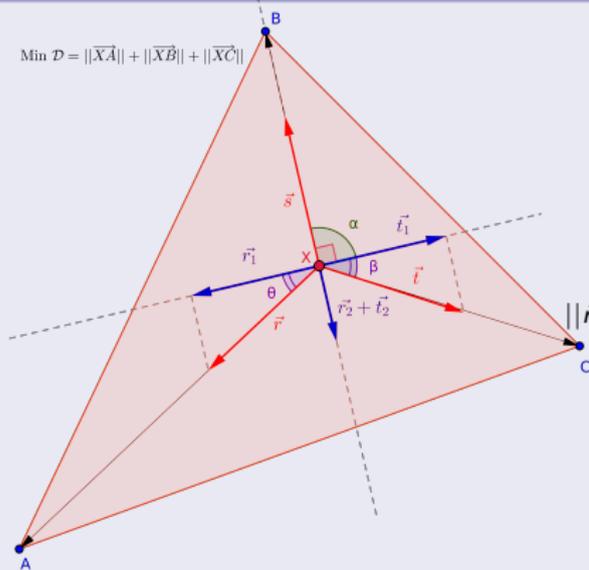
# The History

## Challenge of Fermat in the 17th century

*Given three points in the plane, find a fourth point such that the sum of its distance to the three given points is minimum.*

## Fermat's Challenge as an Optimization Problem

$$\text{Min } \mathcal{D} = \|\vec{XA}\| + \|\vec{XB}\| + \|\vec{XC}\|$$



Three Forces in Equilibrium  
( $0^\circ < \theta, \beta < 90^\circ$ )

$$\begin{aligned}\|\vec{r}_1\| &= \|\vec{t}_1\| \Rightarrow \cos(\theta) = \cos(\beta) \\ &\Rightarrow \theta = \beta\end{aligned}$$

$$\begin{aligned}\|\vec{r}_2 + \vec{t}_2\| &= \|\vec{s}\| \Rightarrow \sin(\theta) + \sin(\beta) = 1 \\ &\Rightarrow \sin(\theta) = \sin(\beta) = \frac{1}{2} \\ &\Rightarrow \theta = \beta = 30^\circ\end{aligned}$$

$$\alpha = 90^\circ + \beta \Rightarrow \alpha = 120^\circ.$$

Now, consider  $p$  given points in  $\mathbb{R}^n$ .

### Steiner Minimal Tree Problem

*Find a minimum tree that spans these points using or not extra points, which are called Steiner points.*

Now, consider  $p$  given points in  $\mathbb{R}^n$ .

### Steiner Minimal Tree Problem

*Find a minimum tree that spans these points using or not extra points, which are called Steiner points.*

This is a very well known problem in combinatorial optimization.

Now, consider  $p$  given points in  $\mathbb{R}^n$ .

### Steiner Minimal Tree Problem

*Find a minimum tree that spans these points using or not extra points, which are called Steiner points.*

This is a very well known problem in combinatorial optimization.

This problem has been shown to be NP-Hard.

Now, consider  $p$  given points in  $\mathbb{R}^n$ .

### Steiner Minimal Tree Problem

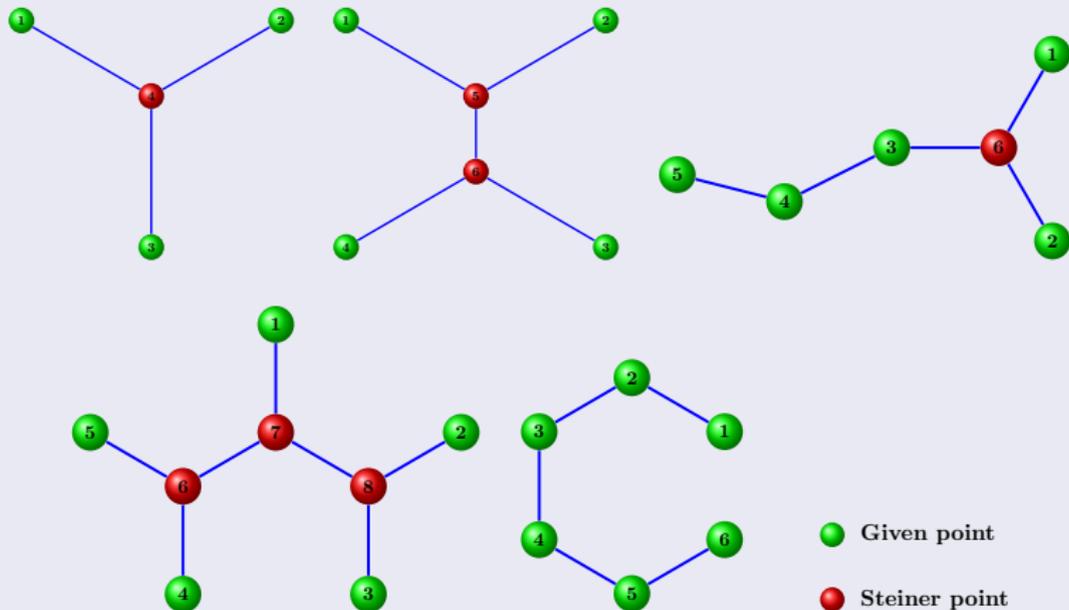
*Find a minimum tree that spans these points using or not extra points, which are called Steiner points.*

This is a very well known problem in combinatorial optimization.

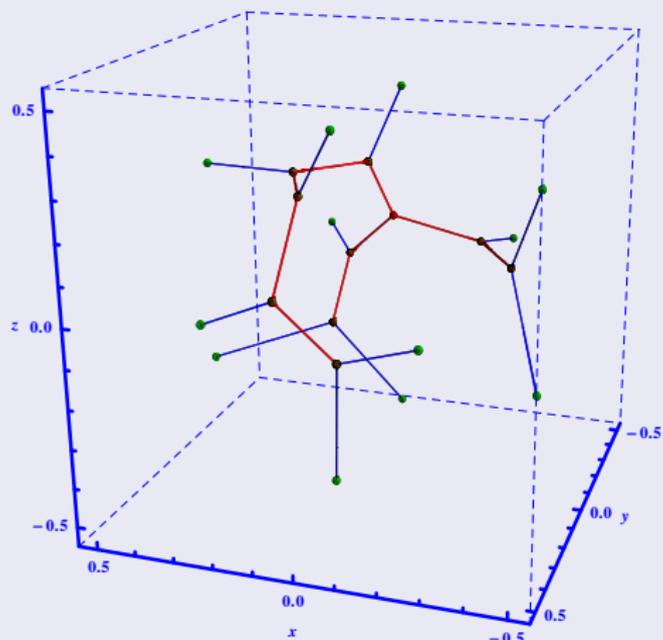
This problem has been shown to be NP-Hard.

All distances are considered to be Euclidean.

## Some examples of Steiner points in $\mathbb{R}^2$



## An example in $\mathbb{R}^3$ : Icosahedron



## Number of Steiner Points

Given  $p$  points  $x^i \in \mathbb{R}^n$ ,  $i = 1, 2, \dots, p$ , the *maximum number of Steiner points* is  $p - 2$ .



## Number of Steiner Points

Given  $p$  points  $x^i \in \mathbb{R}^n$ ,  $i = 1, 2, \dots, p$ , the *maximum number of Steiner points* is  $p - 2$ .

## Degree of Steiner Points

A nondegenerated Steiner point has degree (valence) *equal to 3*.

## Number of Steiner Points

Given  $p$  points  $x^i \in \mathbb{R}^n$ ,  $i = 1, 2, \dots, p$ , the *maximum number of Steiner points* is  $p - 2$ .

## Degree of Steiner Points

A nondegenerated Steiner point has degree (valence) *equal to 3*.

## Steiner Points Edges

The edges emanating from a nondegenerated Steiner point *lie in a plane* and have *mutual angle equal to  $120^\circ$* .

## Steiner Topology

It is a topology that satisfy all the Steiner Tree properties.



## Steiner Topology

It is a topology that satisfy all the Steiner Tree properties.

## Number of Topologies (Gilbert and Pollack)

The total number of different topologies with  $k$  Steiner points is

$$C_{p,k+2} \frac{(p+k-2)!}{k!2^k},$$

where  $p$  is the number of given points in  $\mathbb{R}^n$ .

## Steiner Topology

It is a topology that satisfy all the Steiner Tree properties.

## Number of Topologies (Gilbert and Pollack)

The total number of different topologies with  $k$  Steiner points is

$$C_{p,k+2} \frac{(p+k-2)!}{k!2^k},$$

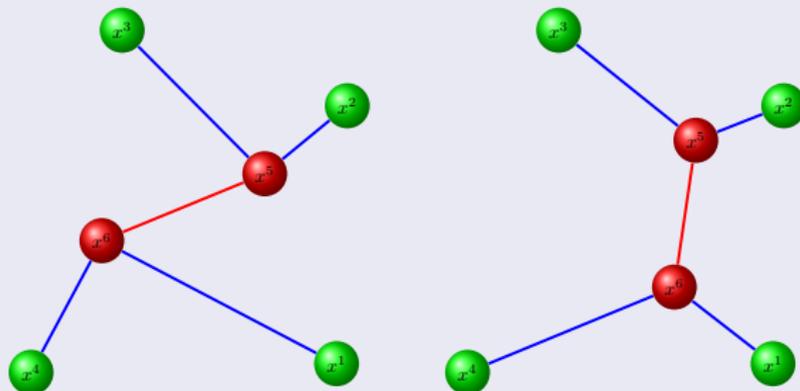
where  $p$  is the number of given points in  $\mathbb{R}^n$ .

## Full Steiner Topologies ( $k = p - 2$ )

The total number of different topologies with  $k = p - 2$  Steiner points is

$$1 \cdot 3 \cdot 5 \cdot 7 \dots (2p - 5) = (2p - 5)!!.$$

## Example of Local Optimization

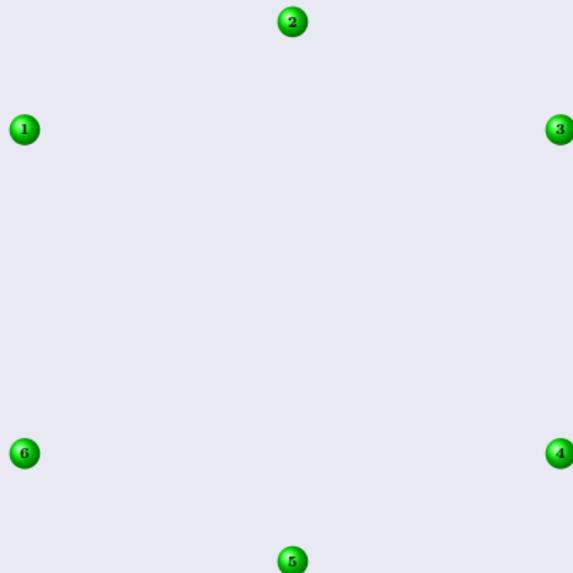


## Finding the best solution...

$$\text{Minimize } \|x^3 - x^5\| + \|x^2 - x^5\| + \|x^5 - x^6\| + \|x^1 - x^6\| + \|x^4 - x^6\|$$
$$\text{subject to } x^5 \text{ and } x^6 \in \mathbb{R}^n.$$

## First Formulation: an example with $p = 6$

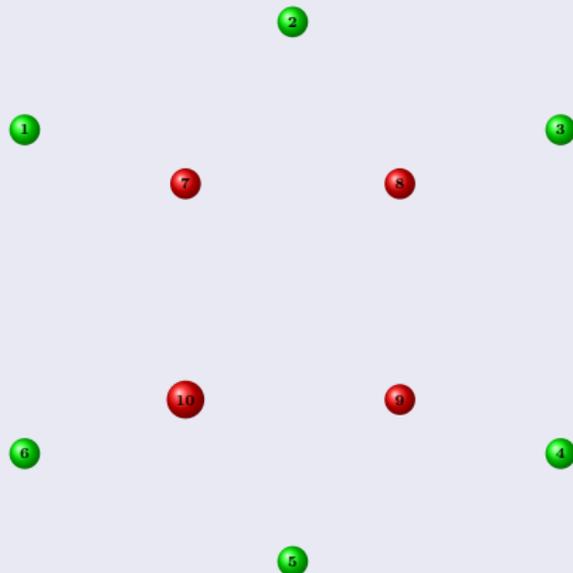
6 given points.



## First Formulation: an example with $p = 6$

6 given points.

4 Steiner points.

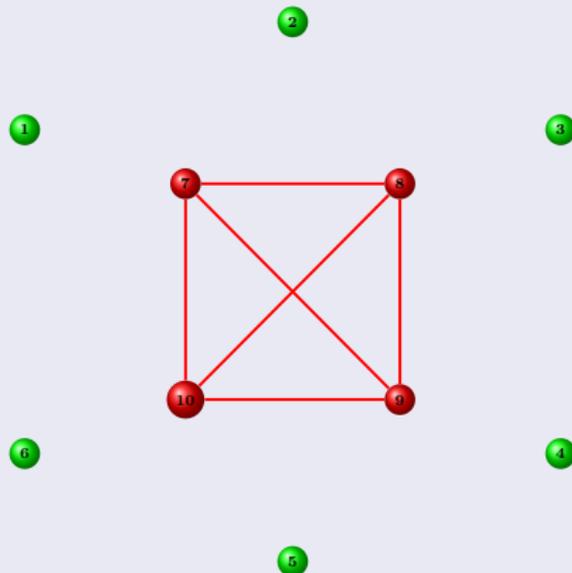


## First Formulation: an example with $p = 6$

6 given points.

4 Steiner points.

All possible edges among  
Steiner points.



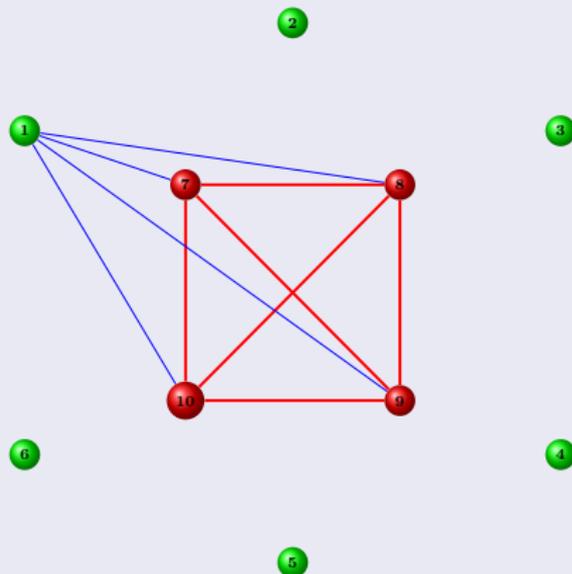
## First Formulation: an example with $p = 6$

6 given points.

4 Steiner points.

All possible edges among Steiner points.

All possible connections between a given point and a Steiner point.



## First Formulation: an example with $p = 6$

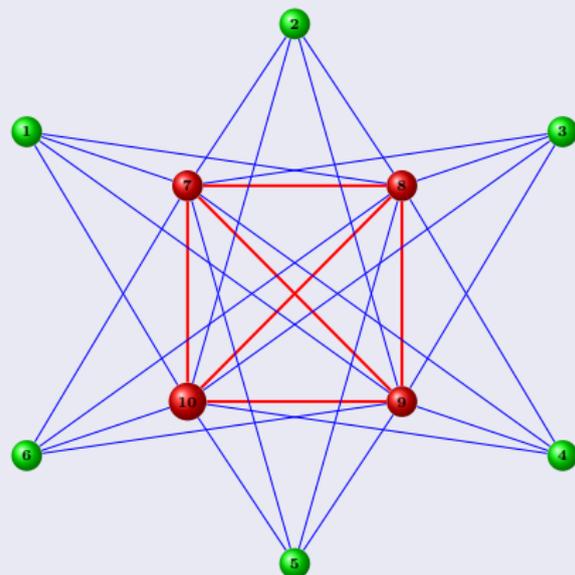
6 given points.

4 Steiner points.

All possible edges among  
Steiner points.

All possible connections between  
a given point and a Steiner  
point.

All possible edges.



## First Formulation: an example with $p = 6$

6 given points.

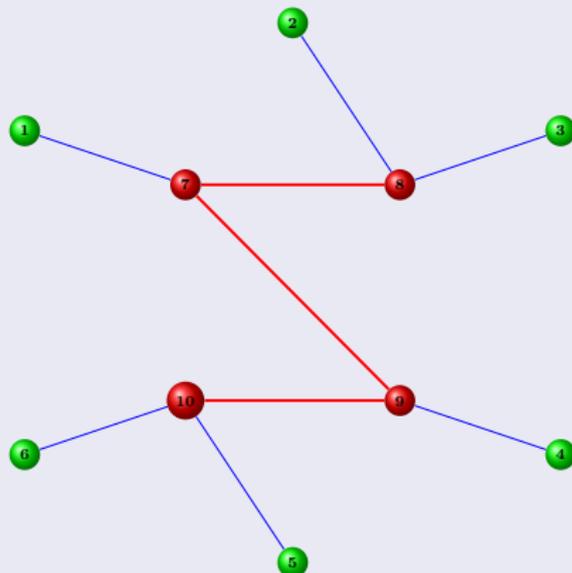
4 Steiner points.

All possible edges among Steiner points.

All possible connections between a given point and a Steiner point.

All possible edges.

An example of a set of possible edges.



Given  $p$  points in  $\mathbb{R}^n$ , we define a especial graph  $G = (V, E)$ .

## First Formulation

$$(P) : \text{Minimize } \sum_{[i,j] \in E} \|x^i - x^j\| y_{ij} \text{ subject to} \quad (1)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P = \{1, 2, \dots, p\}, \quad (2)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (3)$$

$$x^i \in \mathbb{R}^n, \quad i \in S, \quad (4)$$

$$y_{ij} \in \{0, 1\}, \quad [i, j] \in E, \quad (5)$$

where  $\|x^i - x^j\| = \sqrt{\sum_{l=1}^n (x_l^i - x_l^j)^2}$  is the Euclidean distance between  $x^i$  and  $x^j$ .

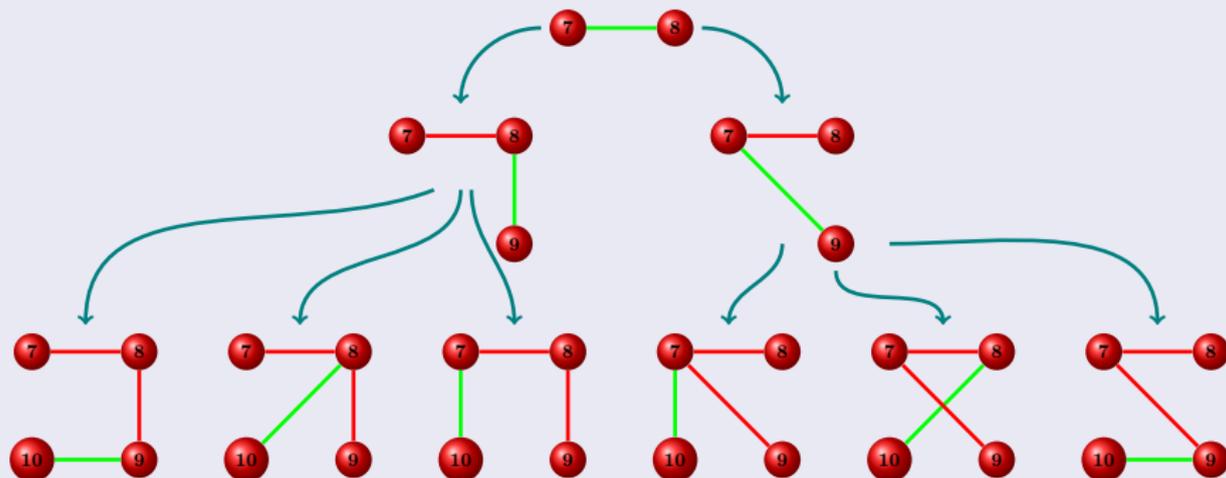
## First Formulation: an example with $p = 6$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p + 1\}$$

$$y_{7,8} = 1$$

$$y_{7,9} + y_{8,9} = 1$$

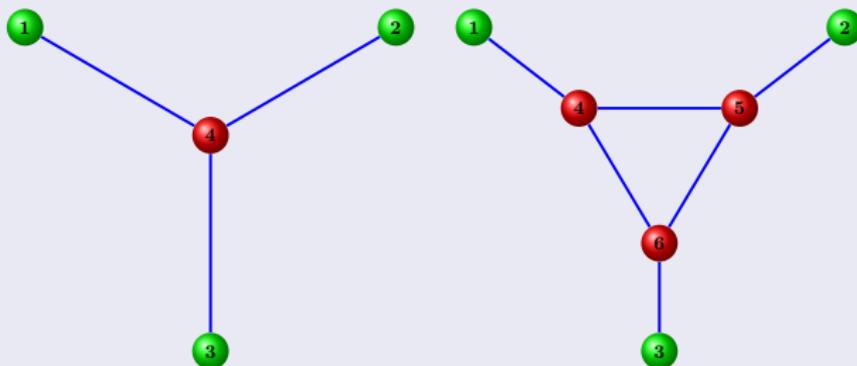
$$y_{7,10} + y_{8,10} + y_{9,10} = 1$$



## First Formulation: another example

If we don't consider

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}$$



## First Formulation (another way to write)

$$(P) : \text{Minimize } \sum_{[i,j] \in E} (t_{ij}^2 - u_{ij}^2) \text{ subject to} \quad (6)$$

$$\|x^i - x^j\| - (t_{ij} + u_{ij}) \leq 0, \quad [i,j] \in E, \quad (7)$$

$$y_{ij} - (t_{ij} - u_{ij}) = 0, \quad [i,j] \in E, \quad (8)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P = \{1, 2, \dots, p\}, \quad (9)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S = \{p+1, \dots, 2p-2\}, \quad (10)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (11)$$

$$x^i \in \mathbb{R}^n, \quad i \in S, \quad (12)$$

$$y_{ij} \in \{0, 1\}, \quad [i,j] \in E. \quad (13)$$

## First Formulation: Lagrangian Relaxation

$$\begin{aligned} \mathcal{L}(x, y, t, u, \alpha, \beta) &= \sum_{[i,j] \in E} (t_{ij}^2 - u_{ij}^2) + \sum_{[i,j] \in E} [||x^i - x^j|| - (t_{ij} + u_{ij})] \alpha_{ij} + \\ &+ \sum_{[i,j] \in E} [y_{ij} - (t_{ij} - u_{ij})] \beta_{ij} \end{aligned}$$

or

$$\begin{aligned} \mathcal{L}(x, y, t, u, \alpha, \beta) &= \sum_{[i,j] \in E} [t_{ij}^2 - u_{ij}^2 - (\alpha_{ij} + \beta_{ij})t_{ij} - (\alpha_{ij} - \beta_{ij})u_{ij}] + \\ &+ \sum_{[i,j] \in E} \alpha_{ij} ||x^i - x^j|| + \sum_{[i,j] \in E} \beta_{ij} y_{ij}, \end{aligned}$$

where

$\alpha_{ij} \geq 0$  is the dual variable associated to constraint (7).

$\beta_{ij} \in \mathbb{R}$  is the dual variable associated to constraint (8).

## First Formulation: Lagrangian Relaxation and The Dual Program

$$\mathcal{D}(\alpha, \beta) = \text{minimum } \{\mathcal{L}(x, y, t, u, \alpha, \beta) \text{ subject to (15) – (20)}\} \quad (14)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P, \quad (15)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S, \quad (16)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (17)$$

$$y_{ij} \in \{0, 1\}, \quad [i, j] \in E, \quad (18)$$

$$0 \leq t_{ij} + u_{ij} \leq M, \quad (19)$$

$$x^i \in R^n, \quad i \in S \quad (20)$$

where  $M = \text{maximum } \{\|x^i - x^j\| \text{ for } 1 \leq i \leq j \leq p\}$ .

## First Formulation: Lagrangian Relaxation and The Dual Program

$$\mathcal{D}(\alpha, \beta) = \text{minimum } \{\mathcal{L}(x, y, t, u, \alpha, \beta) \text{ subject to (15) - (20)}\} \quad (14)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P, \quad (15)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S, \quad (16)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (17)$$

$$y_{ij} \in \{0, 1\}, \quad [i, j] \in E, \quad (18)$$

$$0 \leq t_{ij} + u_{ij} \leq M, \quad (19)$$

$$x^i \in R^n, \quad i \in S \quad (20)$$

where  $M = \text{maximum } \{\|x^i - x^j\| \text{ for } 1 \leq i \leq j \leq p\}$ .

We define

$$\mathcal{D}_1(t, u, \alpha, \beta) = \text{minimum } \left\{ \sum_{[i,j] \in E} [t_{ij}^2 - u_{ij}^2 - (\alpha_{ij} + \beta_{ij})t_{ij} - (\alpha_{ij} - \beta_{ij})u_{ij}] \mid \text{s.t. (19)} \right\},$$

## First Formulation: Lagrangian Relaxation and The Dual Program

$$\mathcal{D}(\alpha, \beta) = \text{minimum } \{ \mathcal{L}(x, y, t, u, \alpha, \beta) \text{ subject to (15) - (20)} \} \quad (14)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P, \quad (15)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S, \quad (16)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (17)$$

$$y_{ij} \in \{0, 1\}, \quad [i, j] \in E, \quad (18)$$

$$0 \leq t_{ij} + u_{ij} \leq M, \quad (19)$$

$$x^i \in R^n, \quad i \in S \quad (20)$$

where  $M = \text{maximum } \{ \|x^i - x^j\| \text{ for } 1 \leq i \leq j \leq p \}$ .

We define

$$\mathcal{D}_2(x, \alpha) = \text{minimum } \left\{ \sum_{[i,j] \in E} \alpha_{ij} \|x^i - x^j\| \mid \text{s.t. (20)} \right\},$$

## First Formulation: Lagrangian Relaxation and The Dual Program

$$\mathcal{D}(\alpha, \beta) = \text{minimum } \{\mathcal{L}(x, y, t, u, \alpha, \beta) \text{ subject to (15) - (20)}\} \quad (14)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P, \quad (15)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S, \quad (16)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (17)$$

$$y_{ij} \in \{0, 1\}, \quad [i, j] \in E, \quad (18)$$

$$0 \leq t_{ij} + u_{ij} \leq M, \quad (19)$$

$$x^i \in R^n, \quad i \in S \quad (20)$$

where  $M = \text{maximum } \{\|x^i - x^j\| \text{ for } 1 \leq i \leq j \leq p\}$ .

We define

$$\mathcal{D}_3(y, \beta) = \text{minimum } \left\{ \sum_{[i,j] \in E} \beta_{ij} y_{ij} \mid \text{s.t. (15) - (18)} \right\},$$

## First Formulation: Lagrangian Relaxation and The Dual Program

$$\mathcal{D}(\alpha, \beta) = \text{minimum } \{\mathcal{L}(x, y, t, u, \alpha, \beta) \text{ subject to (15) - (20)}\} \quad (14)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P, \quad (15)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S, \quad (16)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (17)$$

$$y_{ij} \in \{0, 1\}, \quad [i, j] \in E, \quad (18)$$

$$0 \leq t_{ij} + u_{ij} \leq M, \quad (19)$$

$$x^i \in R^n, \quad i \in S \quad (20)$$

where  $M = \text{maximum } \{\|x^i - x^j\| \text{ for } 1 \leq i \leq j \leq p\}$ .

Thus we can write

$$\mathcal{D}(\alpha, \beta) = \mathcal{D}_1(t, u, \alpha, \beta) + \mathcal{D}_2(x, \alpha) + \mathcal{D}_3(y, \beta).$$

## First Formulation: Lagrangian Relaxation and The Dual Program

$$\mathcal{D}(\alpha, \beta) = \text{minimum } \{\mathcal{L}(x, y, t, u, \alpha, \beta) \text{ subject to (15) – (20)}\} \quad (14)$$

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P, \quad (15)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S, \quad (16)$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p+1\}, \quad (17)$$

$$y_{ij} \in \{0, 1\}, \quad [i, j] \in E, \quad (18)$$

$$0 \leq t_{ij} + u_{ij} \leq M, \quad (19)$$

$$x^i \in R^n, \quad i \in S \quad (20)$$

where  $M = \text{maximum } \{\|x^i - x^j\| \text{ for } 1 \leq i \leq j \leq p\}$ .

The Dual Problem will be

$$\text{Maximize } \mathcal{D}(\alpha, \beta) \text{ subject to} \quad (21)$$

$$\alpha \geq 0, \quad [i, j] \in E, \quad (22)$$

$$\beta \in R, \quad [i, j] \in E. \quad (23)$$

## First Formulation: Lagrangian Relaxation and The Dual Program

The Lagrangian Relaxation and The Dual Program were proposed by

*N. Maculan, P. Michelon and A. E. Xavier, in*

*The Euclidean Steiner problem in  $\mathbb{R}^n$  : A mathematical programming formulation, Annals of Operations Research, vol. 96, pp. 209-220, 2000.*

## The Idea

To improve the enumeration scheme presented by Smith<sup>a</sup>, by the inclusion of **lower bounds** which are obtained from the Dual Problem Solution.

---

<sup>a</sup>W. D. Smith, *How to find Steiner minimal trees in Euclidean d-space*, Algorithmica, vol. 7, pp. 137-177, 1992.

## Second Formulation

$$(P) : \text{Minimize } \sum_{[i,j] \in E} d_{ij} \text{ subject to} \quad (24)$$

$$d_{ij} \geq \|a^i - x^j\| - M(1 - y_{ij}), [i, j] \in E_1, \quad (25)$$

$$d_{ij} \geq \|x^i - x^j\| - M(1 - y_{ij}), [i, j] \in E_2, \quad (26)$$

$$d_{ij} \geq 0, [i, j] \in E \quad (27)$$

$$\sum_{j \in S} y_{ij} = 1, i \in P, \quad (28)$$

$$\sum_{i < j, i \in S} y_{kj} = 1, j \in S - \{p+1\}, \quad (29)$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, j \in S, \quad (30)$$

$$x^i \in \mathbb{R}^n, i \in S, \quad (31)$$

$$y_{ij} \in \{0, 1\}, [i, j] \in E, \quad (32)$$

$$d_{ij} \in \mathbb{R}. \quad (33)$$

We consider

$$\left\{ \begin{array}{l} \|x^i - x^j\| \approx \sqrt{\sum_{l=1}^n (x_l^i - x_l^j)^2} + \lambda^2 \\ M = \text{maximum}\{\|a^i - a^j\| \text{ for } 1 \leq i < j \leq p\}, \\ E_1 = \{[i, j] | i \in P, j \in S\}, E_2 = \{[i, j] | i \in S, j \in S\} \text{ e } E = E_1 \cup E_2 \end{array} \right.$$

## Second Formulation (First Property)

If  $\bar{x}^j \in R^n$ ,  $j \in S$  and  $\bar{y}_{ij} \in \{0, 1\}$ ,  $[i, j] \in E$  is an optimal solution, then

$$d_{ij} = \|a^i - \bar{x}^j\| \geq 0 \text{ or } d_{ij} = 0, \text{ for all } [i, j] \in E_1 \text{ and}$$

$$d_{ij} = \|\bar{x}^i - \bar{x}^j\| \geq 0 \text{ or } d_{ij} = 0, \text{ for all } [i, j] \in E_2.$$

## Second Formulation (Second Property)

$y_{ij} \in \{0, 1\}$ ,  $[i, j] \in E$  is associated with a full Steiner Topology if, and only if, the following equations are satisfied:

$$\sum_{j \in S} y_{ij} = 1, \quad i \in P,$$

$$\sum_{k < j, k \in S} y_{kj} = 1, \quad j \in S - \{p + 1\},$$

$$\sum_{i \in P} y_{ij} + \sum_{k < j, k \in S} y_{kj} + \sum_{k > j, k \in S} y_{jk} = 3, \quad j \in S,$$

## Note that...

When we consider

$$\|x^i - x^j\| \approx \sqrt{\sum_{l=1}^n (x_l^i - x_l^j)^2 + \lambda^2},$$

error propagations may happen.

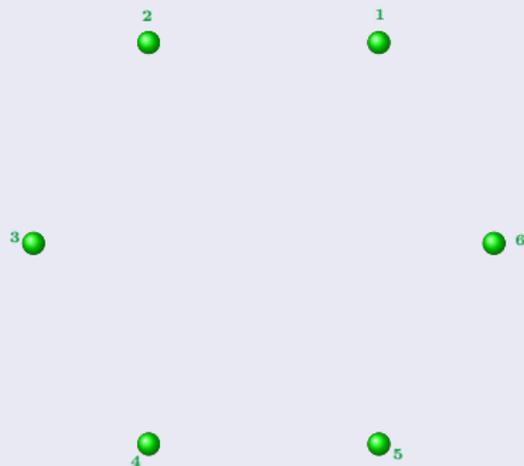
## Note that...

When we consider

$$\|x^i - x^j\| \approx \sqrt{\sum_{l=1}^n (x_l^i - x_l^j)^2 + \lambda^2},$$

error propagations may happen.

## Example: Regular Hexagon



6 given points.

Each given point is in a vertex of a Regular Hexagon.

Each side of the Hexagon is equal to 1.

# MINLP: Formulations for the Euclidean Steiner Problem

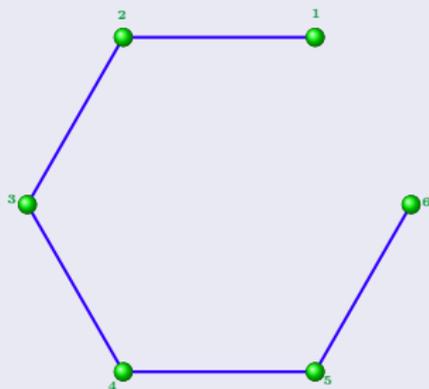
## Note that...

When we consider

$$\|x^i - x^j\| \approx \sqrt{\sum_{l=1}^n (x_l^i - x_l^j)^2 + \lambda^2},$$

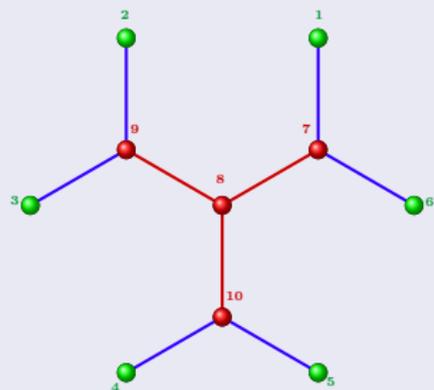
error propagations may happen.

## Example: Regular Hexagon



Objective Function: 5

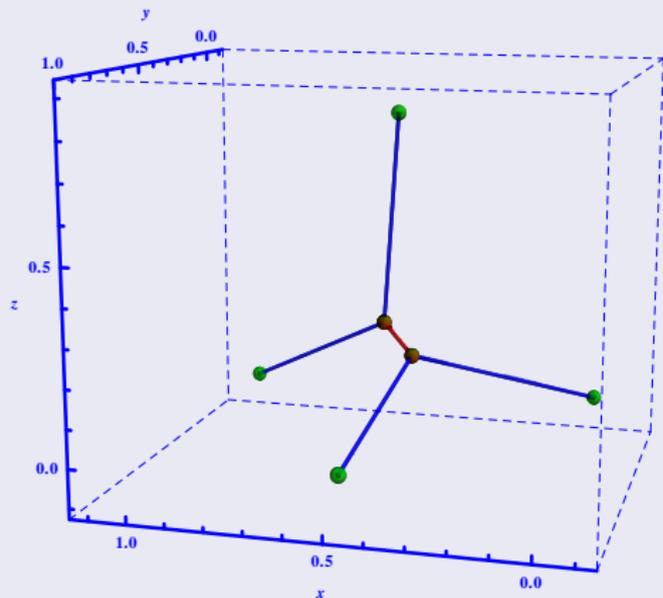
$$\lambda^2 = 10^{-8}$$



Objective Function:  $5.196 = 3\sqrt{3}$

$$\lambda^2 = 10^{-6}$$

## Second Formulation: One Solution for a Tetrahedron



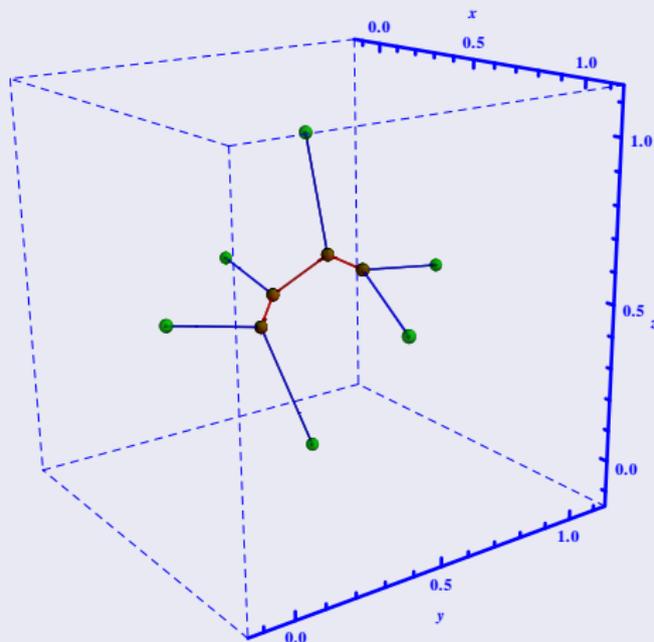
Number of Points (Green): 4

Number of Steiner Points (Red): 2

Objective Function: 2.43911

Execution Time: 3.27 s

## Second Formulation: One Solution for an Octahedron



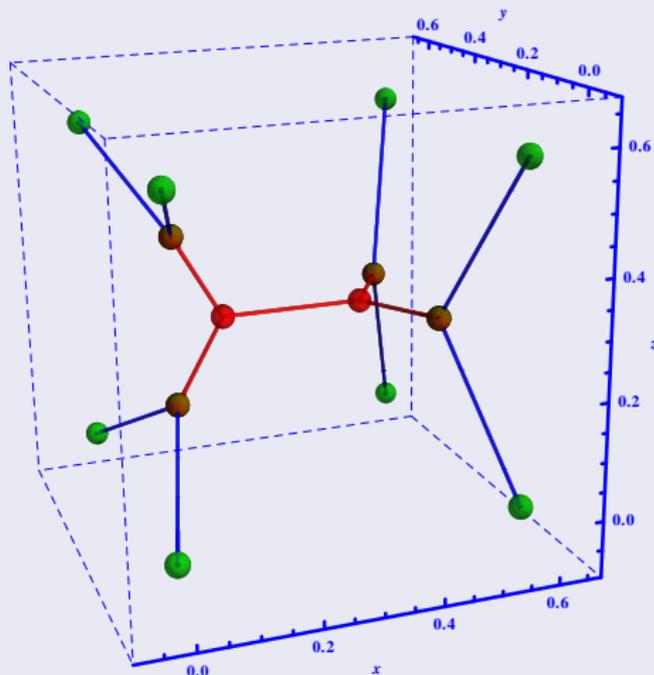
Number of Points (Green): 6

Number of Steiner Points (Red): 4

Objective Function: 2.86801

Execution Time: 2.22 min

## Second Formulation: One Solution for a Cube



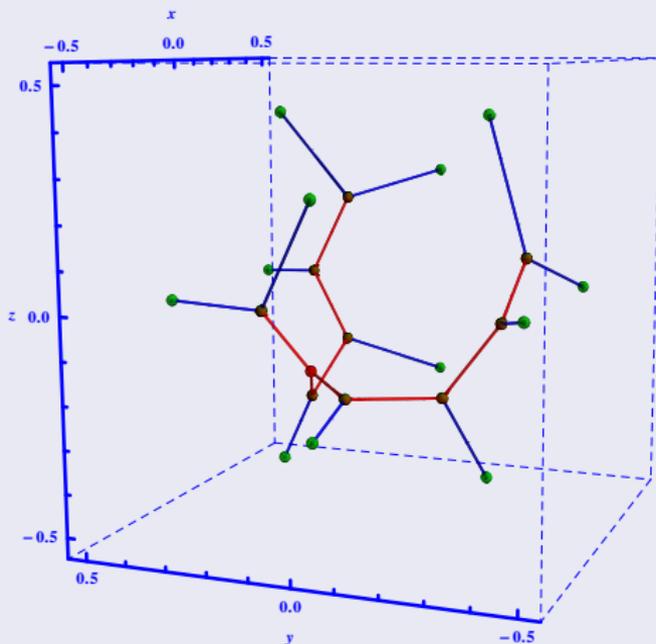
Number of Points (Green): 8

Number of Steiner Points (Red): 6

Objective Function: 3.57735

Execution Time: 3 h

## Second Formulation: One Solution for an Icosahedron



Number of Points (Green): 12  
Number of Steiner Points (Red): 10  
Objective Function: 4.90531  
Execution Time: 48 h (not finished).

Thank you!