

Method of Orienting Lines for Minimizing a Sum of Euclidean Norms

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Minimizing a Sum of Euclidean Norms

$$\min_t \left(\mathcal{F}(t_1, \dots, t_k) = \sum_{i=1}^{k-1} \|p_i(t_i) - p_{i+1}(t_{i+1})\| \right), \quad (1)$$

where $t_i \geq a_i > 0$, $p_i : \mathbf{R}^1 \rightarrow \mathbf{R}^2$, ($i = 1, \dots, k$) are linear, $k \geq 3$ and $\|\cdot\|$ is an Euclidean norm in $\mathbf{R}^2 \implies$

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- How can we find exact solutions of Problem (1)/(1*)?

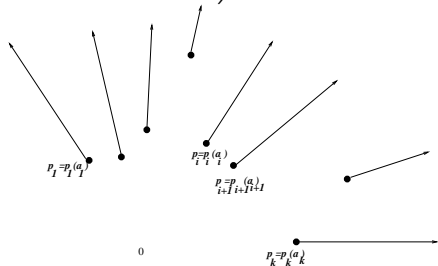
Geometrical Form of Problem (1)

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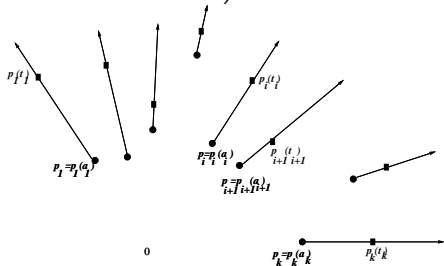
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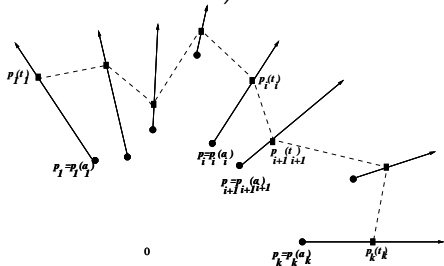
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Optimal Control Problem

Some optimal control problems can be stated in the form

$$(P) \quad \min_{x,u} \int_{t_0}^{t_f} F(t, x(t), u(t)) dt$$

subject to

$$\dot{x}(t) = f(t, x(t), u(t)), \quad t \in [t_0, t_f]$$

$$\alpha_1(t) \geq x(t) \geq \alpha_2(t), \quad x(t_0) = x_0, \quad x(t_f) = x_f$$

$$g(t, u(t)) \geq 0, \quad t \in [t_0, t_f].$$

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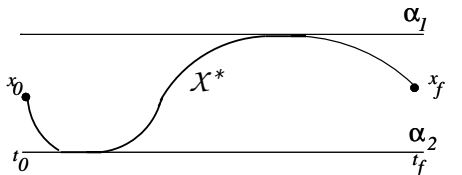
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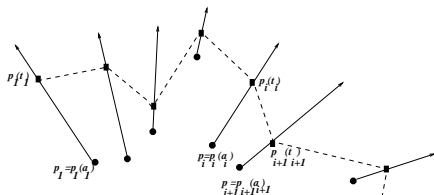
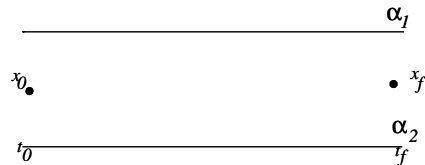
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New concepts: Final curve, Orienting curve \implies

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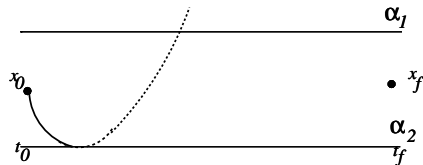
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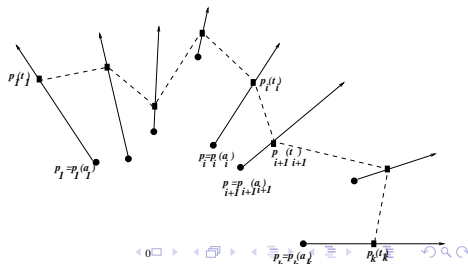
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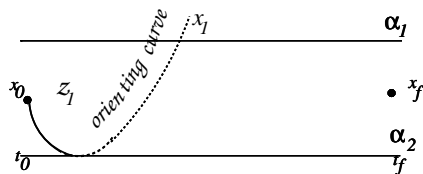
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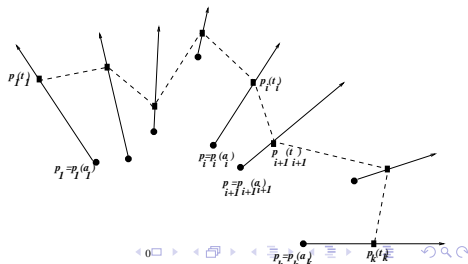
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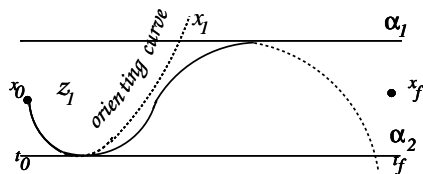
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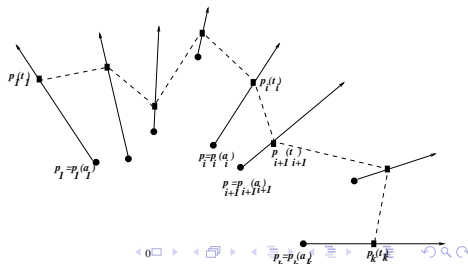
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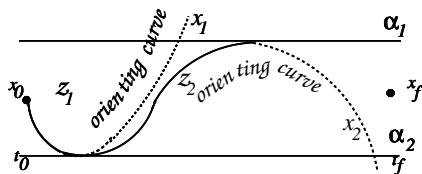
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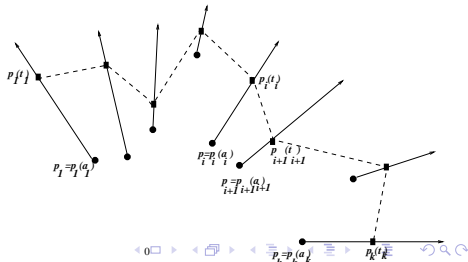
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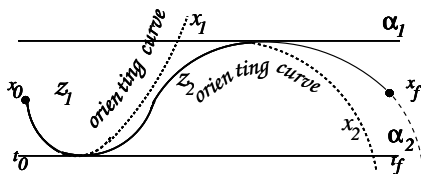
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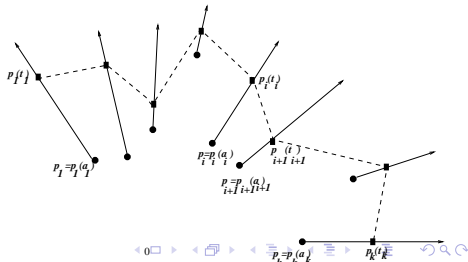
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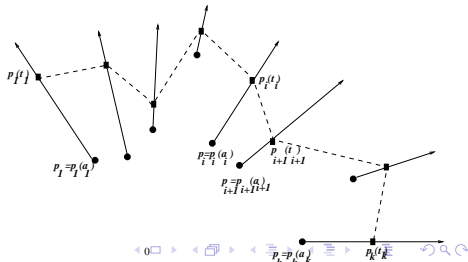
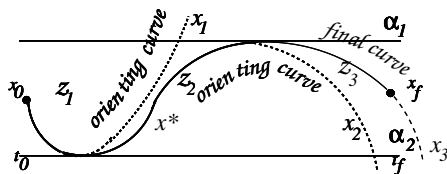
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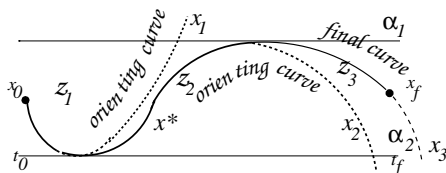
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- Optimal control problem (P):
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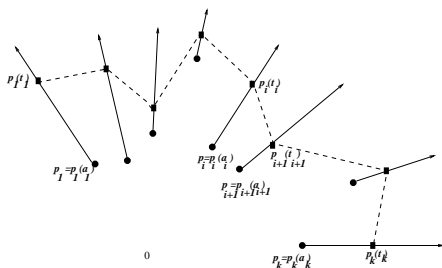
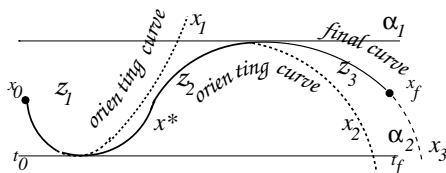
Method of Orienting Curves

- Optimal control problem (P):
Final curve
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optimal solution consists of parts of
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- Minimizing a Sum of Euclidean Norms (1): Can it be solved exactly by the idea of the Method of Orienting Curves above?

Final curve \rightarrow Final line? Orienting curve \rightarrow Orienting line?

Difficulty: First and final points, boundaries α_1, α_2 are unknown!!!



Method of Orienting Lines

Take $\|p_m(a_m)\| = \max_{1 \leq i \leq k} \{\|p_i(a_i)\|\}$. Then $p_m(a_m)$ belongs to the solution $Q(a)$ of Problem (1).

Let α be

the sector of the circle

radius $\max\{\|p_i\| : m \leq i \leq k\}$ centered at 0,

between two rays $\overrightarrow{Op_m}$ and $\overrightarrow{Op_k}$ contains $p_m(a_m), \dots, p_i(a_k)$.

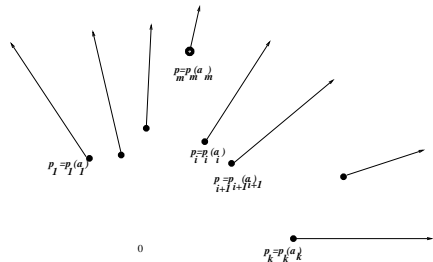
\implies The path formed by the solution $Q(a)$ of Problem (1) is the **shortest path inside the domain** formed by the polyline $\alpha_1 := p_m(a_m) \dots p_i(a_k)$ and $\alpha_1 := \text{arc}_\alpha$ **with unknown final point.**

Method of Orienting Lines

→ Boundaries:

$\alpha_1 := p_m(a_m) \dots p_i(a_k), \alpha_1 := \text{arc}_{\alpha_1}$.

→ We start from $p_m(a_m)$ to construct final line and orienting lines

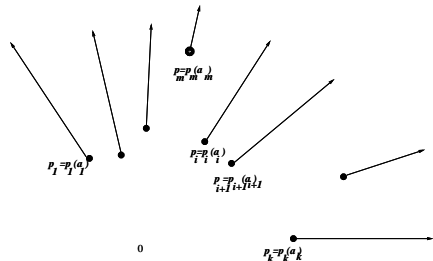


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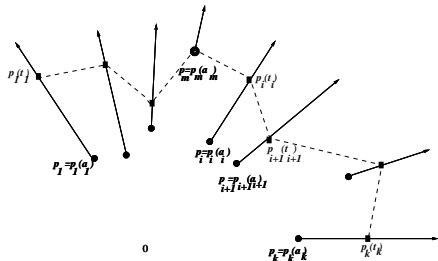


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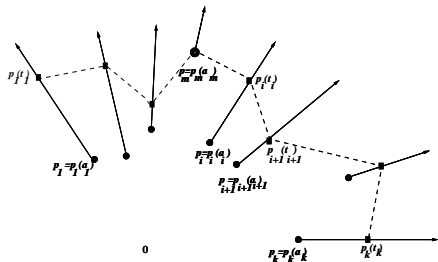


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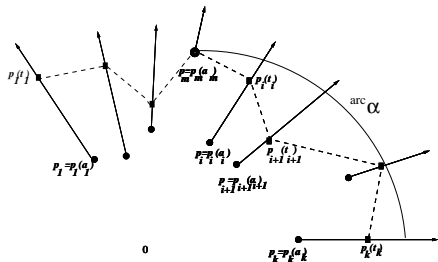


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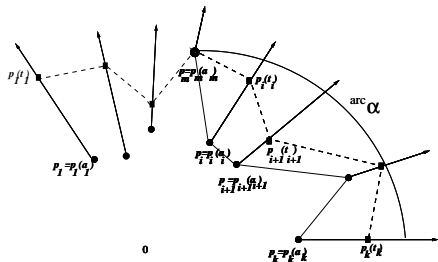


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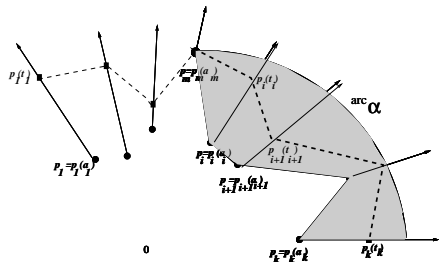


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Final Line and Orienting Line

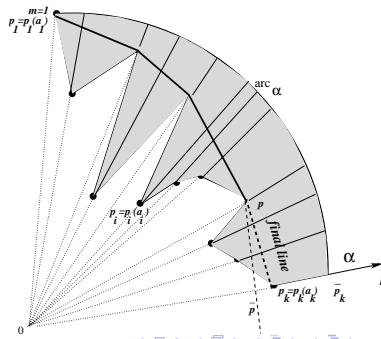
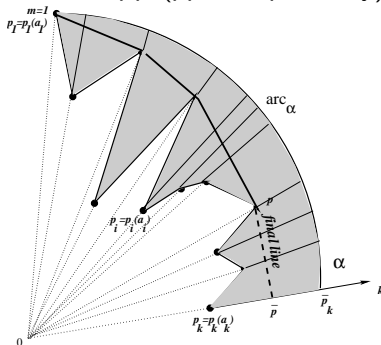
Let $p \in Q(a)$ (solution of Problem (1)), $\bar{p} \in 0p_k$ such that

$$p\bar{p} \perp 0p_k$$

z is on or right of $p\bar{p} \quad \forall z \in Q(a) \setminus \{p_k\}$, z between p and p_{k-1}

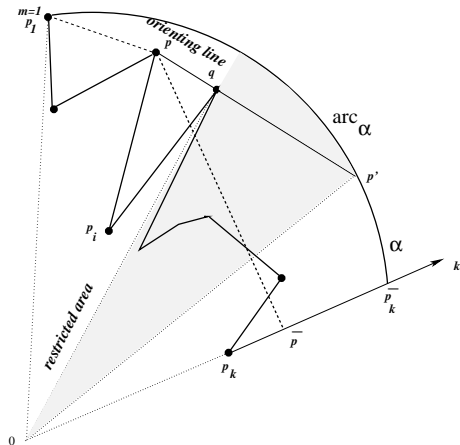
p_k is right (on or left, respectively) of $p\bar{p}$

$\implies p\bar{p}$ (pp_k , respectively) is a *final line* through p .



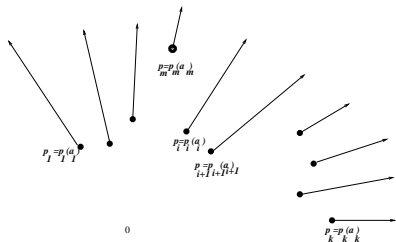
Final Line and Orienting Line

$p, q \in Q(a) \setminus \{p_k\}$ such that q
 between p and p_{k-1} .
 z is on or right of $\vec{pq} \forall z \in Q(a)$
 between p and p_k ,
 $\implies pq$ is an *orienting line* through p .



Exact Algorithm

We now determine $Q(a) = Q^R(t^*)$ (a part of solution of Problem (1)).



Exact solution of (1) consists of parts of orienting lines and final lines.

1. Begin at p_m . Set $l = 1$. Then,

$q_1 = p_m$.

2. Consider q_l .

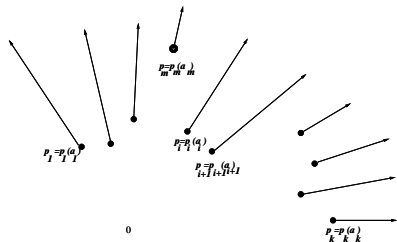
If there is a final line through q_l go to **4** else, there is an orienting line through q_l , go to **3**.

3. Let $q_l p_{k_l}^*$ be an orienting line with $p_{k_l}^*$ as its transfer point. Then, $p_{k_l}^* \in Q^R(t^*)$. Set $q_{l+1} = p_{k_l}^*$ and $l = l + 1$, go to **2**.

4. Let $q_l q$ be the final line, where $q \in \vec{p}_k$. Then $Q^R(t^*)$ includes $\{q_1, \dots, q_l, q\}$.

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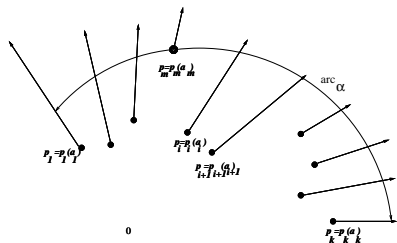
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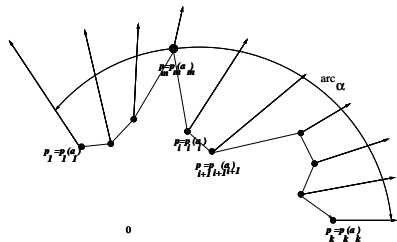
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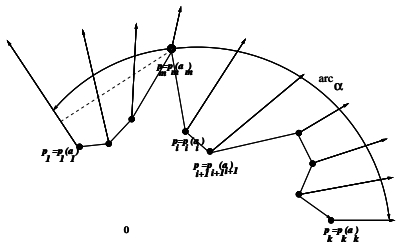
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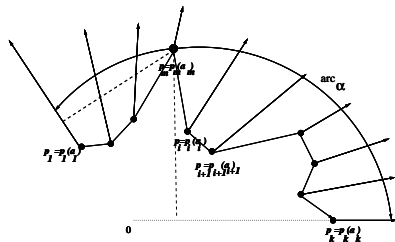
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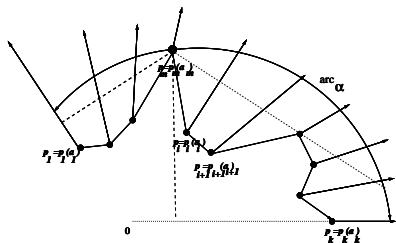
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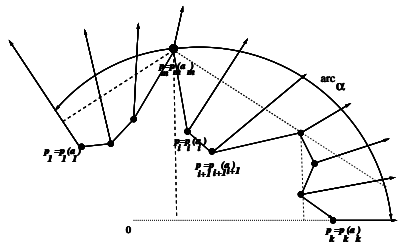
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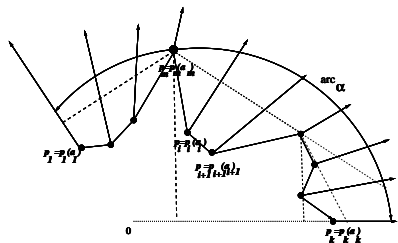
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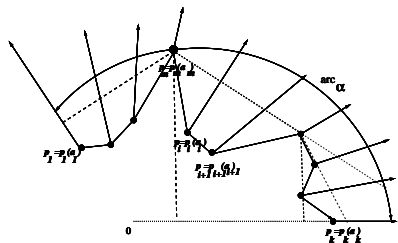
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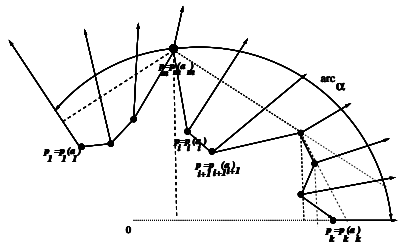
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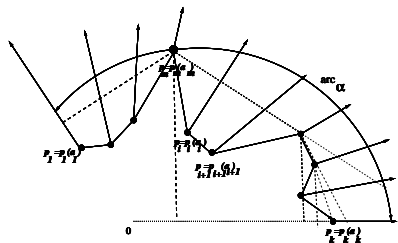
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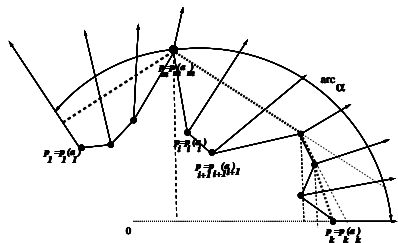
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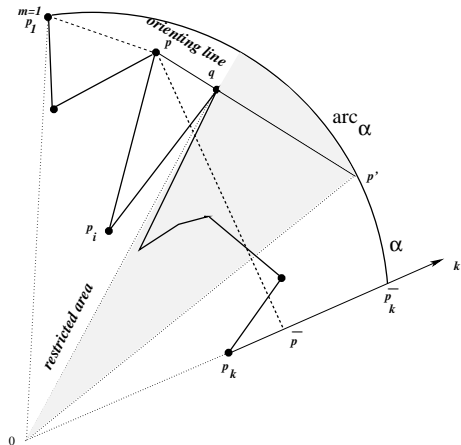
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Restricted Areas

It helps to determine quickly if $q_i p$ is orienting line or not:



Application Example

$$\inf_t \mathcal{F}(t) \quad (2)$$

where

$$\begin{aligned} \mathcal{F}(t) = & \sqrt{(2t_7 - 30t_6)^2 + 36t_6^2} + \sqrt{(30t_6 - 53t_5)^2 + (6t_6 - 16t_5)^2} \\ & + \sqrt{(53t_5 - 4t_4)^2 + (16t_5 - 5t_4)^2} + \sqrt{(4t_4 - 3t_3)^2 + (5t_4 - 7t_3)^2} \\ & + \sqrt{(3t_3 - t_2)^2 + (7t_3 - 21t_2)^2} + \sqrt{(-5t_1 - t_2)^2 + (5t_1 - 21t_2)^2} \end{aligned}$$

and $t_7 \geq a_7 = 1$, $t_6 \geq a_6 = 1.5$, $t_5 \geq a_5 = 1$, $t_4 \geq a_4 = 14$, $t_3 \geq a_3 = 3$, $t_2 \geq a_2 = 5$ and $t_1 \geq a_1 = 3.5$.

Application Example

	Approximation algorithms	Our exact algorithm
	-(2) written as a second order cone program -interior-point methods	-Final lines, orienting lines -Restricted area
$\mathcal{F}(t^*)$	≈ 209.636414	$= 56/5 + 1512/265 + 2814/53$ $+ (\sqrt{646594} + \sqrt{1530400}) / 33$ $+ 55\sqrt{2}$
t_1^*	≈ 9.99998	$= 10$
t_2^*	≈ 5	$= 5$
t_3^*	≈ 11.9829	$= 395/33$
t_4^*	≈ 14	$= 14$
t_5^*	≈ 1.0566	$= 56/53$
t_6^*	≈ 1.8666	$= 28/15$
t_7^*	≈ 28.0002	$= 28$

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This method does not rely on triangulation and graph tools.
- *Open question*: Can minimizing a sum of Euclidean norms in higher dimensions with nonlinear p_i be solved *exactly* in the same manner? \implies Exact solution consists of parts of orienting Xs and final Xs???

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