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# Nonholonomic Distance to Polygonal Obstacles for a Car-Like Robot of Polygonal Shape

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joint work with

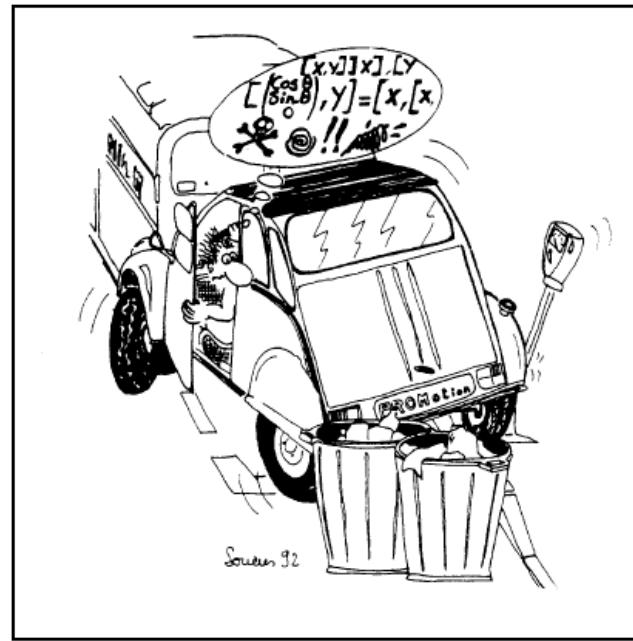
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Philippe Souères

LAAS - CNRS, Toulouse

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...otherwise known as....



## The Optimal Car Crashing Problem

# Outline

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- Introduction
- Shortest paths for the Reeds and Shepp car
- Problem decomposition
- Distance definition
- Necessary conditions for optimality
- Family refinement

# Introduction

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## motivation

- motion planning algorithms rely on the obstacles distance computation; the more accurate the distance, the more efficient the planning
- euclidean distance is not appropriate for nonholonomic robots
- Reeds & Shepp shortest paths induce a metric in the configuration space [Laumond and Souères, 1993]

## definition

- the **distance** from a car-like robot configuration to an obstacle is the length of the shortest feasible path bringing one point on the robot boundary in contact with the obstacle

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## related work

- sufficient family of shortest feasible paths linking two configurations  
[Reeds and Shepp, 1990; Sussmann and Tang, 1991; Boissonnat *et al*, 1992; Souères and Laumond, 1996]
- shortest paths to polygonal obstacles for a point robot  
[Vendittelli and Laumond, 1996]
- shortest paths to polygonal obstacles for a polygonal robot  
[Mirtich *et al*, 1996; Vendittelli *et al*, 1999]

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## contribution

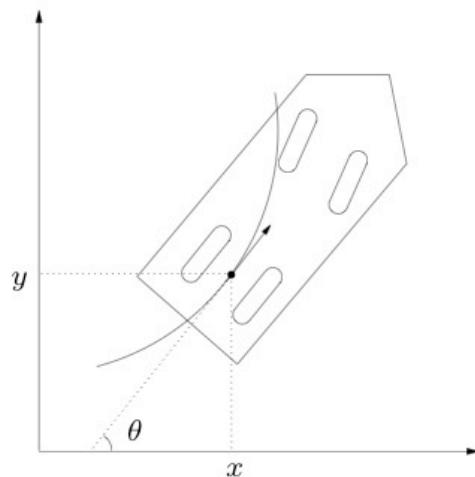
- characterization of the shortest paths to polygonal obstacles for a car-like robot of polygonal shape
- definition of an algorithm for distance computation

## tool

- Pontryagin Maximum Principle + transversality conditions

# Shortest paths for the Reeds & Shepp car

## RS model



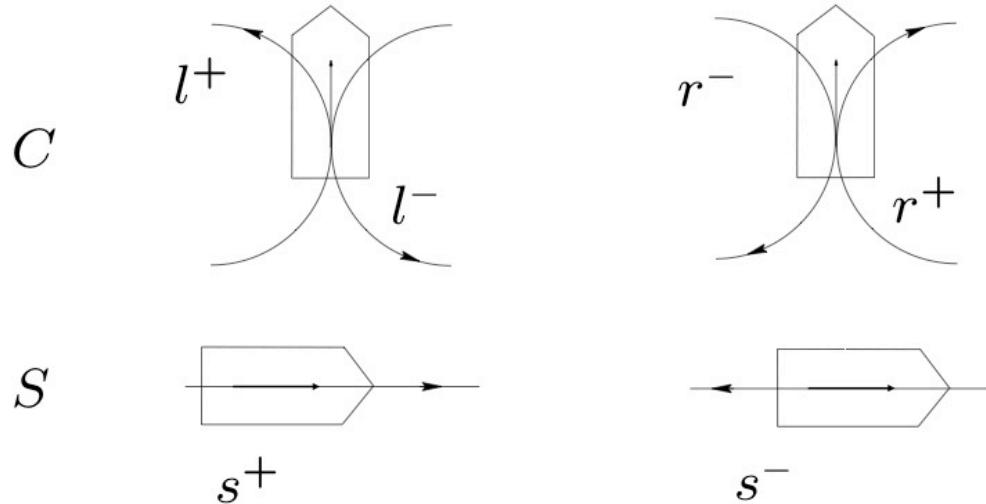
$$\dot{\xi} = f(\xi, u) = g_1(\xi)u_1 + g_2(\xi)u_2$$

$$\xi(t) = (x(t), y(t), \theta(t))$$

$$g_1(\xi) = \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix} \quad g_2(\xi) = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$|u_1(t)| = 1 \quad |u_2(t)| \leq 1$$

## notation

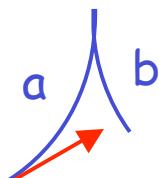


$C, S$  elementary path types

$|_a^{+(-)}, r_a^{+(-)}, s_a^{+(-)}$  elementary paths of length a

example:

path  $|_a^+|_b^-$



is of type  $C|C$

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cost functional

$$J = \int_{t_i}^{t_f} dt$$

Hamiltonian

$$H = \langle \psi, f \rangle = \psi_1 \cos \theta u_1 + \psi_2 \sin \theta u_1 + \psi_3 u_2$$

with

$$\dot{\psi}(t) = -\frac{\partial H}{\partial \xi}(\psi(t), \xi(t), u(t))$$

PMP: consider an admissible  $u(t)$  and the corresponding trajectory  $\xi(t)$ : a necessary condition for  $\xi(t)$  to be optimal is that there exist a nontrivial  $\psi(t)$  and a constant  $\psi_0 \leq 0$  s.t.

$$\begin{aligned} -\psi_0 &= \langle \psi(t), g_1(\xi(t)) \rangle u_1 + \langle \psi(t), g_2(\xi(t)) \rangle u_2 \\ &= \max_{v=(v_1, v_2) \in U} (\langle \psi(t), g_1(\xi(t)) \rangle v_1 + \langle \psi(t), g_2(\xi(t)) \rangle v_2) \end{aligned}$$

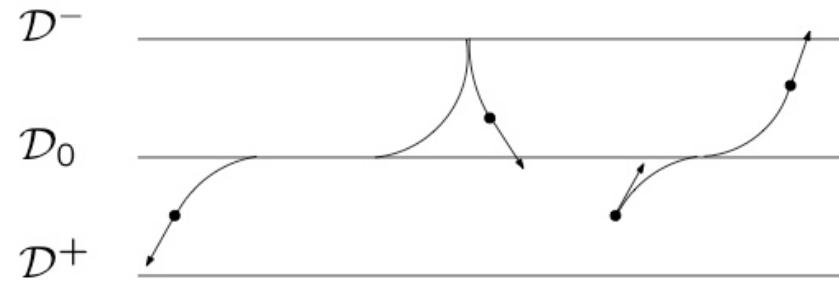
$$t \in [t_i, t_f]$$

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the sufficient family of optimal paths contains

A paths  $C|C|\dots|C$  with  $\text{length}(C) \leq \pi$  for any  $C$

B paths lying between two parallel lines  $\mathcal{D}^+$  and  $\mathcal{D}^-$



straight line segments and inflection points are on  $\mathcal{D}_0$   
cusps lie perpendicularly on  $\mathcal{D}^+$  and  $\mathcal{D}^-$

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with

$$\mathcal{D}_0 : \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_0) = 0$$

$$\mathcal{D}^+ : \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_0) + \psi_0 = 0$$

$$\mathcal{D}^- : \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_0) - \psi_0 = 0$$

- $\psi_1$  and  $\psi_2$  constant
- $\psi_1/\psi_2 = \text{slope of } \mathcal{D}^+, \mathcal{D}^- \text{ and } \mathcal{D}_0$
- $\psi_3(t) = \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_0) \quad (\Rightarrow \mathcal{D}_0 : \psi_3(t) = 0)$

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## the sufficient family $\mathcal{F}$ of optimal paths

$$(I) \quad l_a^+ l_b^- l_e^+ \text{ or } r_a^+ r_b^- r_e^+ \\ 0 \leq a \leq \pi, 0 \leq b \leq \pi, 0 \leq e \leq \pi$$

$$(II)(III) \quad C_a | C_b C_e \text{ or } C_a C_b | C_e \\ 0 \leq a \leq b, 0 \leq e \leq b, 0 \leq b \leq \pi/2$$

$$(IV) \quad C_a C_b | C_b C_e \\ 0 \leq a \leq b, 0 \leq e \leq b, 0 < b \leq \pi/2$$

$$(V) \quad C_a | C_b C_b | C_e \quad \text{a, b, e: RS path parameters}$$

$$(VI) \quad C_a | C_{\pi/2} S_b C_{\pi/2} | C_e \\ 0 \leq a \leq \pi/2, 0 \leq b, 0 \leq e < \pi/2$$

$$(VII)(VIII) \quad C_a | C_{\pi/2} S_b C_e \text{ or } C_a S_b C_{\pi/2} | C_e \\ 0 \leq a \leq \pi, 0 \leq b, 0 \leq e \leq \pi/2$$

$$(IX) \quad C_a S_b C_e \\ 0 \leq a \leq \pi/2, 0 \leq b, 0 \leq e \leq \pi/2$$

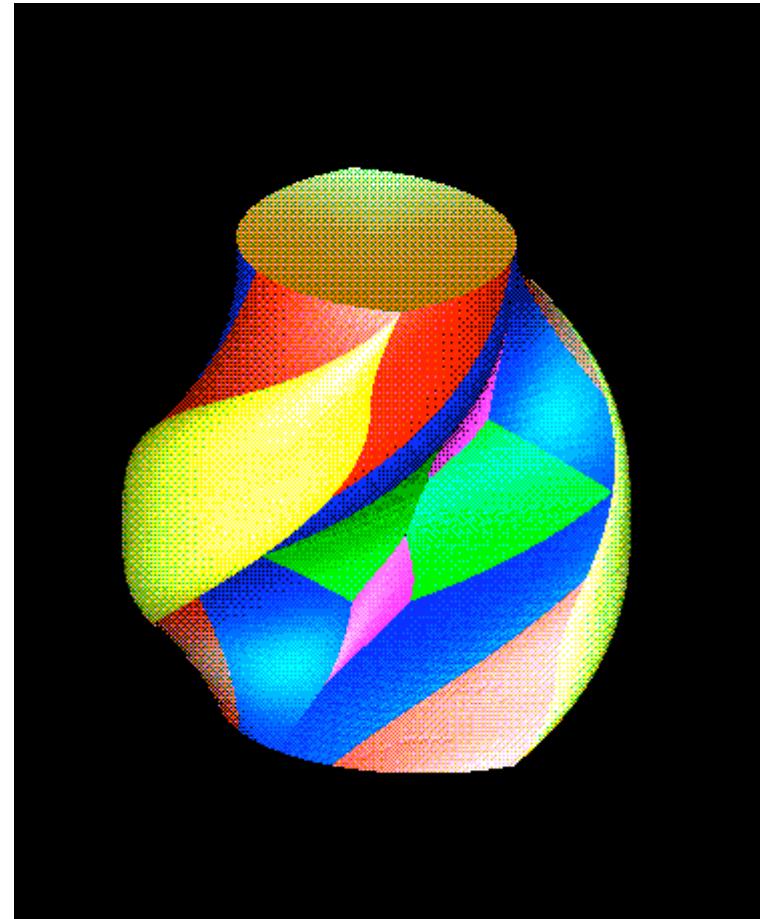
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to every path  $p_i$  is associated a smooth function

$$\begin{pmatrix} x \\ y \\ \theta \end{pmatrix} = W_i(a, b, e) = \begin{pmatrix} X_i(a, b, e) \\ Y_i(a, b, e) \\ \Theta_i(a, b, e) \end{pmatrix}$$

mapping the three RS parameters into the configuration space  
[Mirtich *et al*, 1996]

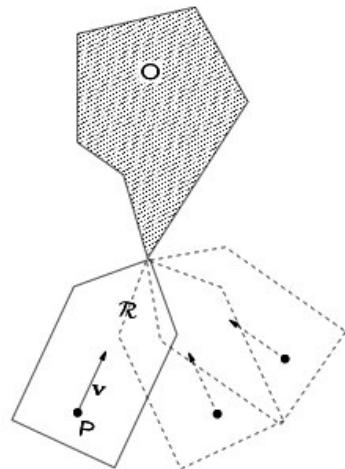
- it is possible to build a “sphere” (the RS ball) of radius  $\ell$  around a given robot configuration  $\xi_0$  [Laumond and Souères, 1993]
- points on the sphere are reachable by paths of length  $\ell$
- points inside the sphere are closer than  $\ell$  to  $\xi_0$



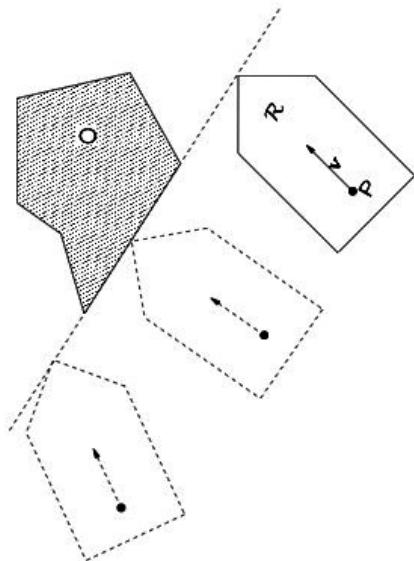
# Problem decomposition

it is sufficient to solve the three subproblems of bringing in contact

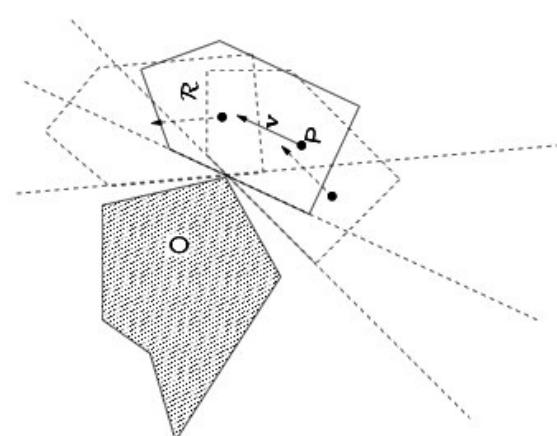
a robot vertex with  
an obstacle vertex  
(VV)



a robot vertex with  
the line supporting  
an obstacle edge  
(VE)



the line supporting  
a robot edge with  
an obstacle vertex  
(EV)



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## idea

- a “contact manifold” can be associated to each sub problem and each vertex/edge of the robot/obstacles

$$M_c(x, y, \theta) = \{(x, y, \theta) \mid \text{there is a VV, VE, EV contact}\}$$

- the distance is determined by the RS ball of minimum radius tangent to one of the  $M_c$ 's

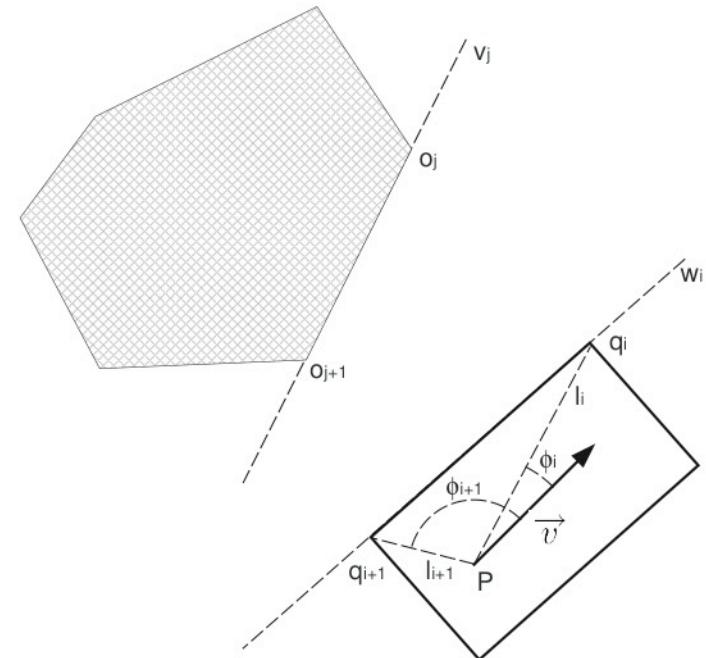
## distance function definition

given the optimal solutions to each of the three sub problems

$$\begin{aligned} 1. \quad L^{VV} : \quad \mathbb{R}^4 &\rightarrow \mathbb{R} \\ (q_i, o_j) &\rightarrow L^{VV}(q_i, o_j) \end{aligned}$$

$$\begin{aligned} 2. \quad L^{VE} : \quad \mathbb{R}^4 &\rightarrow \mathbb{R} \\ (q_i, v_j) &\rightarrow L^{VE}(q_i, v_j) \end{aligned}$$

$$\begin{aligned} 3. \quad L^{EV} : \quad \mathbb{R}^4 &\rightarrow \mathbb{R} \\ (w_i, o_j) &\rightarrow L^{EV}(w_i, o_j) \end{aligned}$$



the distance function is

$$d() : \mathbb{R}^8 \rightarrow \mathbb{R}$$

$$d() = \min \left\{ \min_{i,j} L^{VV}(q_i, o_j), \min_{i,j} L^{VE}(q_i, v_j), \min_{i,j} L^{EV}(w_i, o_j) \right\}$$

## proposed approach to the solution of the three sub problems

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- definition of the contact manifold

$M_c(x, y, \theta) \rightarrow$  constraint on the final state  $\chi(x_f, y_f, \theta_f) = 0$

- for each path  $p_i \in \mathcal{F}$ , closed-form solution in the RS parameter space

$$\chi(W_i(a, b, e)) = 0 \quad (*)$$

the problem is underconstrained: the number of constraints are < 3

VV

one-dimensional  $M_c \rightarrow (*)$  is a system of 2 equations with 3 unknowns

VE and EV

two-dimensional  $M_c \rightarrow (*)$  defines 1 equation in 3 unknowns

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to make the problem "square"

- type A paths
  - geometric properties of the RS ball
    - the boundary of the C|C|C domain coincides with the level curve  $l=|\theta|$
- type B paths
  - transversality conditions

$$\psi_f = \frac{\partial \chi(\xi_f)}{\partial \xi_f} \zeta$$

where

$$\psi_f = (\psi_1(t_f), \psi_2(t_f), \psi_3(t_f))^T$$

$$\xi_f = (x(t_f), y(t_f), \theta(t_f))^T$$

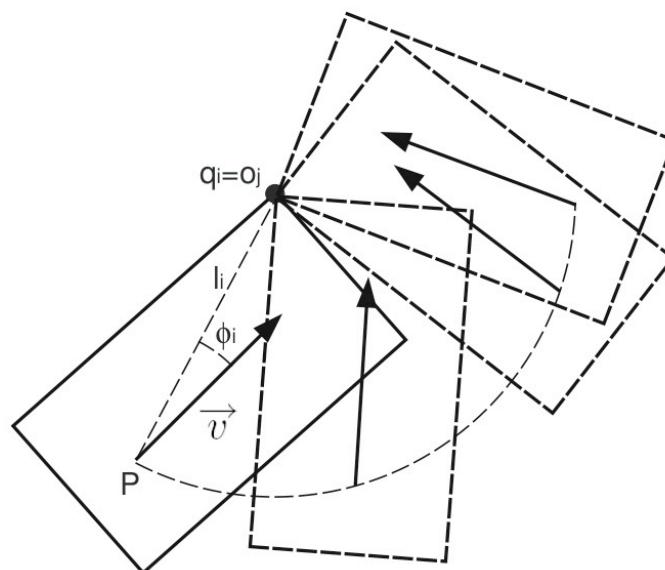
# Vertex-Vertex

## ■ contact manifold

$$\mathcal{M}_c(x, y, \theta) = \{(x, y, \theta) | q_i = o_j\}$$

$$\begin{cases} q_i = (x + l_i \cos(\theta + \phi), y + l_i \sin(\theta + \phi)) \\ o_j = (o_{j_x}, o_{j_y}) \end{cases}$$

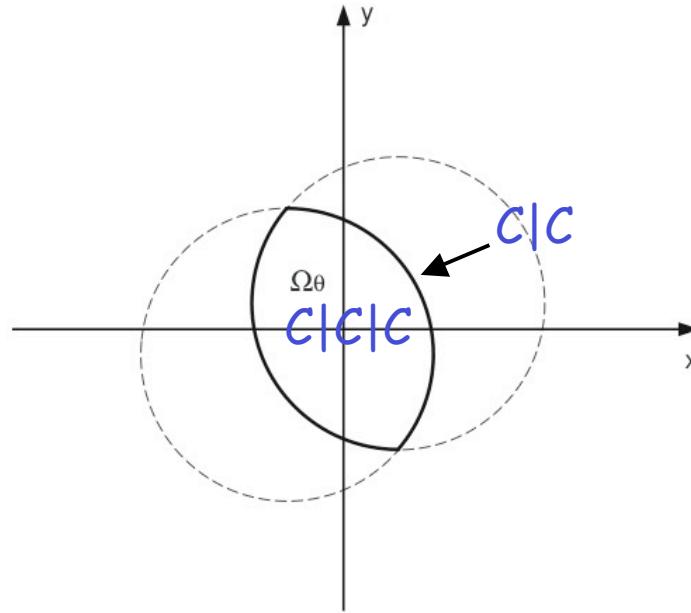
$$\chi(x_f, y_f, \theta_f) = \begin{pmatrix} x_f - o_{j_x} + l_i \cos(\theta_f + \phi_i) \\ y_f - o_{j_y} + l_i \sin(\theta_f + \phi_i) \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$



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- type A paths

if a path of type A is optimal for the VV problem then it is a C|C path



~~>  $\chi(W_{p_k}(0, b, e)) = 0$  is square

- type B paths

if a path of type B is optimal for the VV problem then the corresponding line  $D_0$  passes through the point obstacle  $o_j$

$$\psi_f = \frac{\partial \chi(\xi_f)}{\partial \xi_f} \zeta \quad \left\{ \begin{array}{lcl} \frac{\partial \chi(\xi_f)}{\partial \xi_f} & = & \begin{pmatrix} 1 & 0 & -l_i \sin(\theta_f + \phi_i) \\ 0 & 1 & l_i \cos(\theta_f + \phi_i) \end{pmatrix} \\ \psi_1 & = & \zeta_1 \\ \psi_2 & = & \zeta_2 \\ \psi_3(t_f) & = & -l_i \sin(\theta_f + \phi_i) \zeta_1 + l_i \cos(\theta_f + \phi_i) \zeta_2 \end{array} \right.$$

$$\psi_3(t) = \psi_1 y(t) - \psi_2 x(t) + \psi_3(t_0)$$

$$\psi_3(t_0) = -\psi_1 o_{j_y} + \psi_2 o_{j_x}$$

$$\psi_3(t) = \psi_1(y - o_{j_y}) - \psi_2(x - o_{j_x}) = 0$$

- to each path  $p_k \in \mathcal{F}$  we associate a map  $VV_{p_k}(q_i, o_j)$  which solves VV for the couple  $(q_i, o_j)$  using  $p_k$ ; it is determined by the solution of

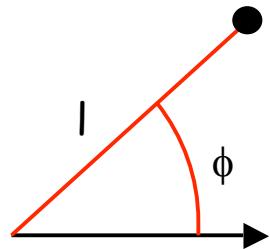
$$\begin{cases} \chi(W_k(a, b, e)) = 0 \\ \psi_1 \cdot (\bar{Y}_k(a, b, e) - o_{j_y}) - \psi_2 \cdot (\bar{X}_k(a, b, e) - o_{j_x}) = 0 \end{cases}$$

where  $\bar{Y}_k(a, b, e)$  and  $\bar{X}_k(a, b, e)$  represent the position of the robot on the line  $\mathcal{D}_0$  computed via  $W_k(a, b, e)$

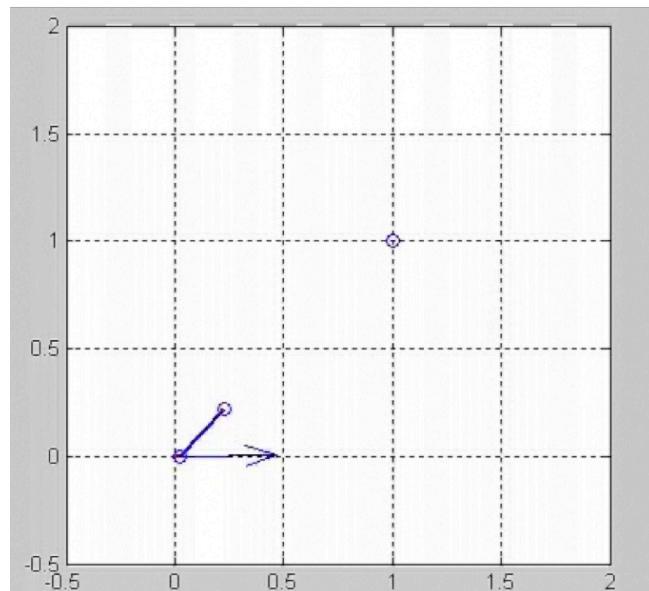
- the optimal solution to the problem VV is

$$L^{VV}(q_i, o_j) = \min_{p_k \in \mathcal{F}} \text{length } VV_{p_k}(q_i, o_j)$$

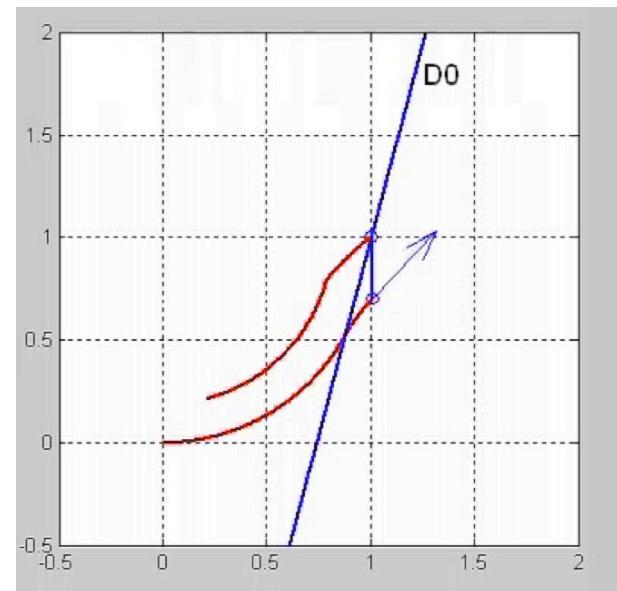
## VV - example 1



$l = 0.3, \phi = \pi/4, \text{ obstacle in } (1,1)$



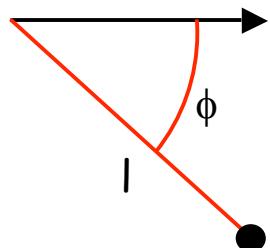
start



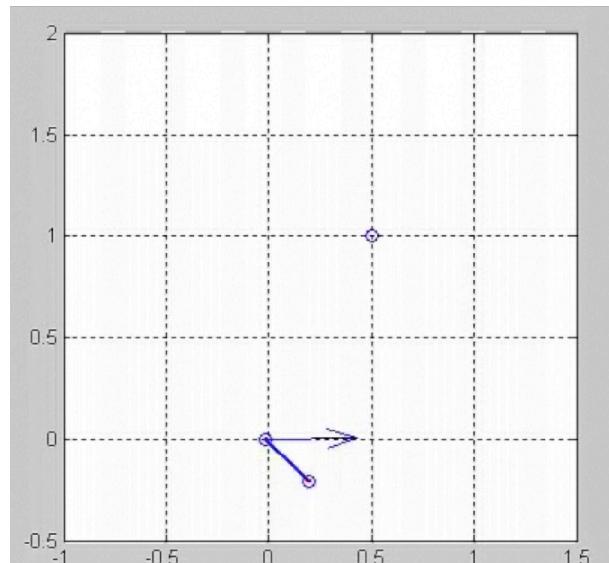
end

$C_a C_b$  type

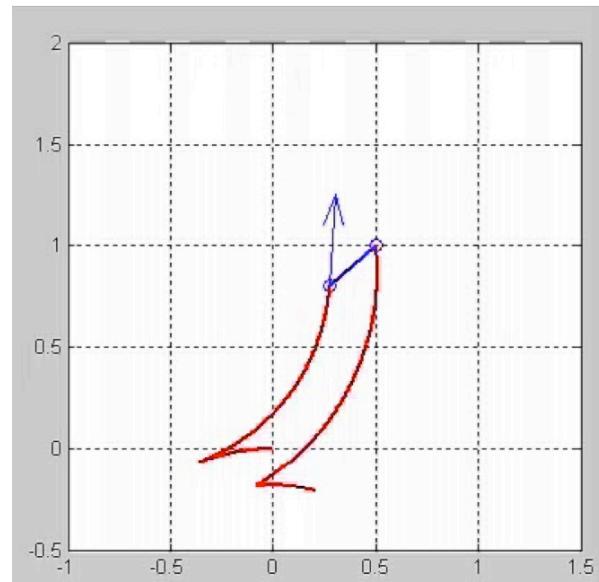
## VV - example 2



$l = 0.3, \phi = -\pi/4, \text{ obstacle in } (0.5, 1)$



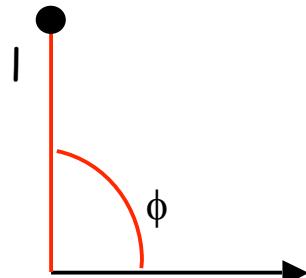
start



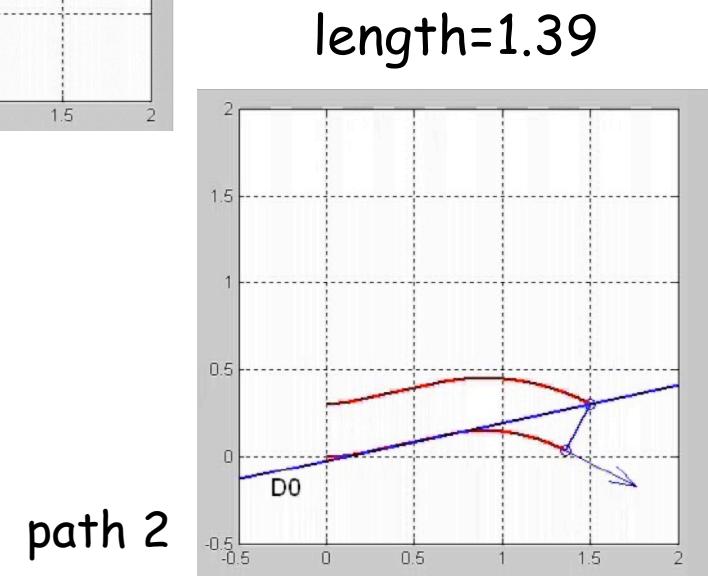
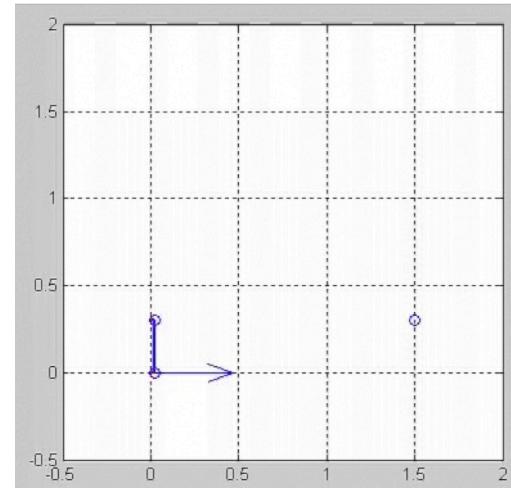
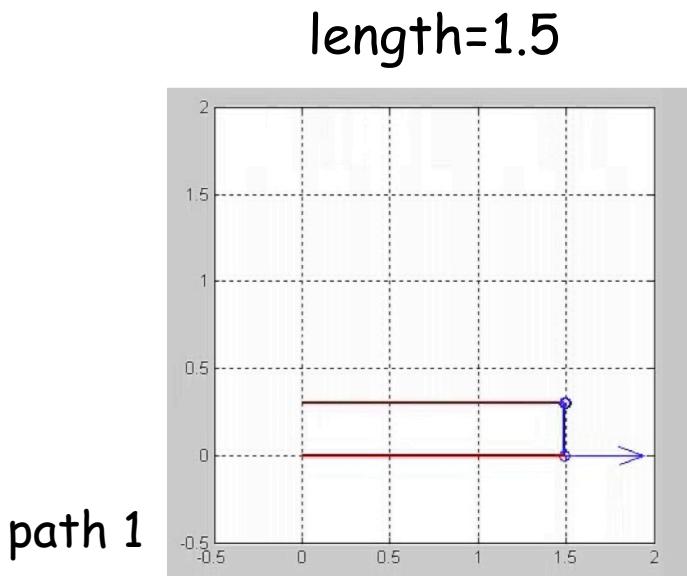
end

$C_a | C_b$  type

## VV - example 3



$l = 0.3, \phi = \pi/2, \text{ obstacle in } (1.5, 0.3)$



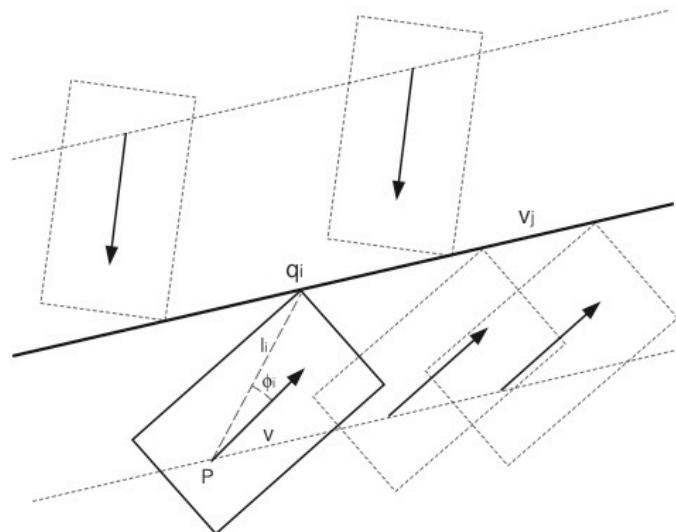
# Vertex-Edge

## ■ contact manifold

$$\mathcal{M}_c(x, y, \theta) = \{(x, y, \theta) \mid q_i \in v_j\}$$

$$\begin{aligned} q_i &= (x + l_i \cos(\theta + \phi), y + l_i \sin(\theta + \phi)) \\ v_j : \bar{y} &= m_i \bar{x} + n_j \end{aligned}$$

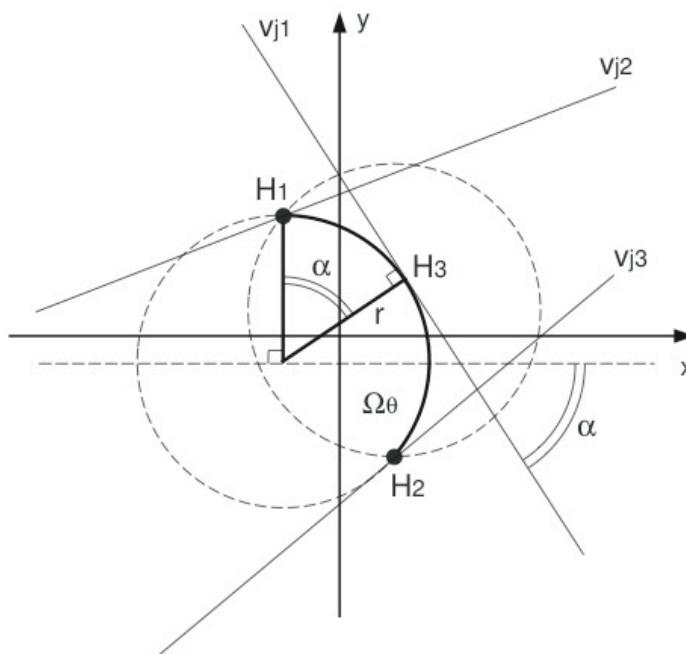
$$\chi(x, y, \theta) = y - m_j x - n_j - l_i m_j \cos(\theta + \phi_i) + l_i \sin(\theta + \phi_i) = 0$$



- type A paths

if a path of type A is optimal for the VV problem then it is a C|C path  
but  $\chi(W_{p_k}(0, b, e)) = 0$  is now underspecified

an "ad hoc" analysis is needed to obtain a square system of equations



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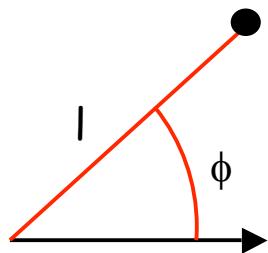
- type B paths

if a path of type B is optimal for the VE problem then

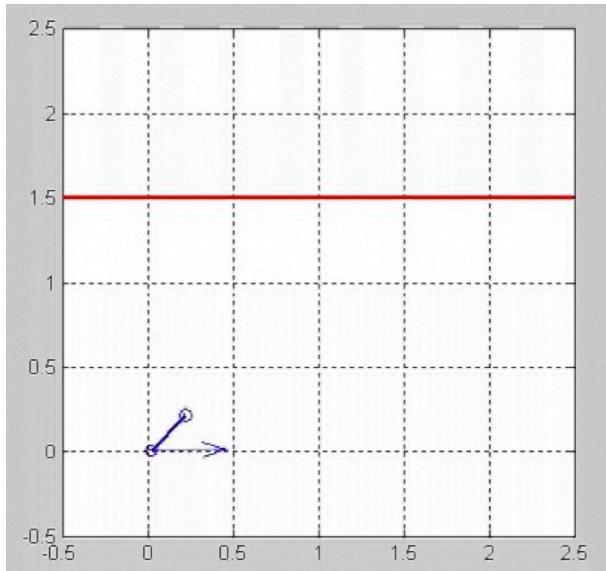
1. the corresponding line  $D_0$  is perpendicular to the line  $v_j$
2. the contact point lies at the intersection between  $D_0$  and  $v_j$

- the contact constraint and the above two necessary conditions allow to define a system of three equations in the three unknowns  $(a, b, c)$
- a solution is sought for each path in the sufficient family  $\mathcal{F}$
- as in the VV case, the optimal solution to the VE problem is obtained by choosing the path of minimum length among all the paths in  $\mathcal{F}$

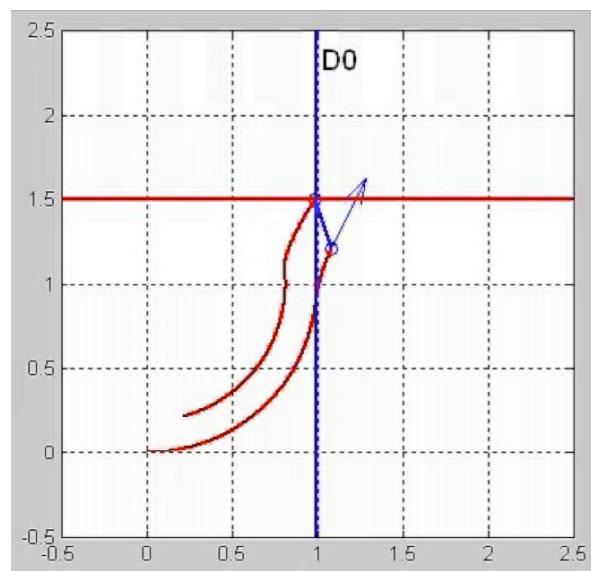
## VE - example



$$l = 0.3, \phi = \pi/4, v_j: \gamma = 1.5$$



start



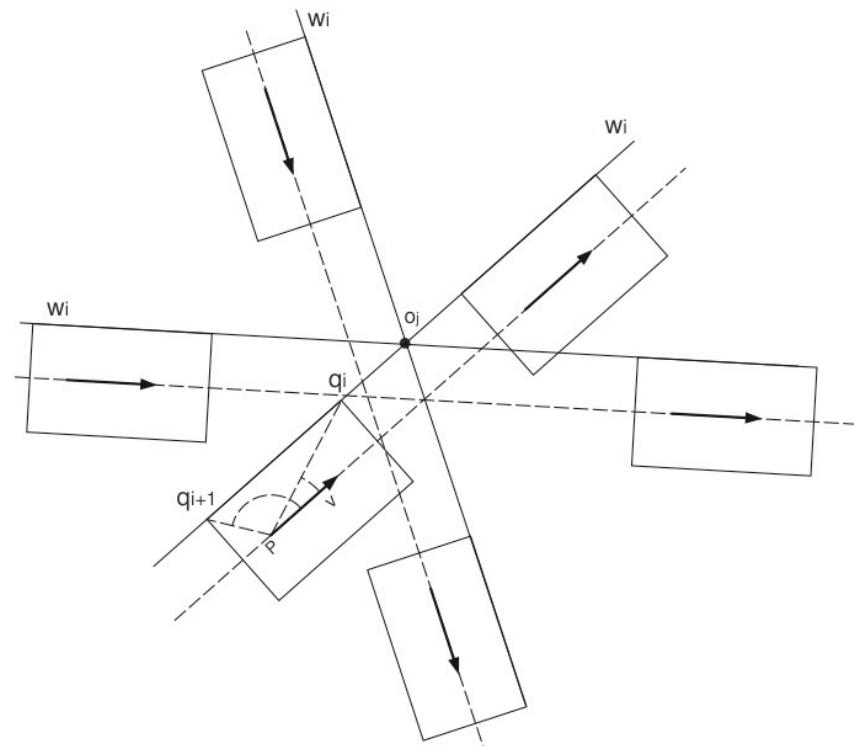
end

$C_a C_b$  type

# Edge-Vertex

- contact manifold

$$\mathcal{M}_c(x, y, \theta) = \{(x, y, \theta) | o_j \in w_i\}$$



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- type A paths

if a path of type A is optimal for the VV problem then it is a C|C path  
but  $\chi(W_{p_k}(0, b, e)) = 0$  is underspecified

the same analysis of VE apply to obtain a square system of equations

- type B paths

if a path of type B is optimal for the EV problem then

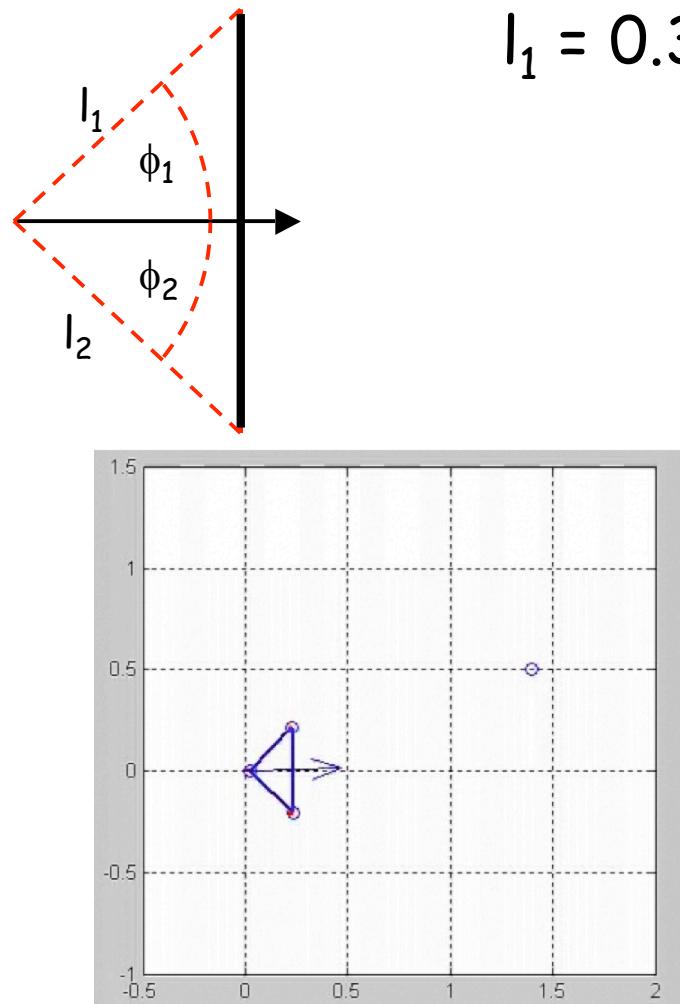
1. the corresponding line  $D_0$  is perpendicular to the edge  $w_j$

at the end of the path

2. the contact point lies at the intersection between  $D_0$  and  $w_j$

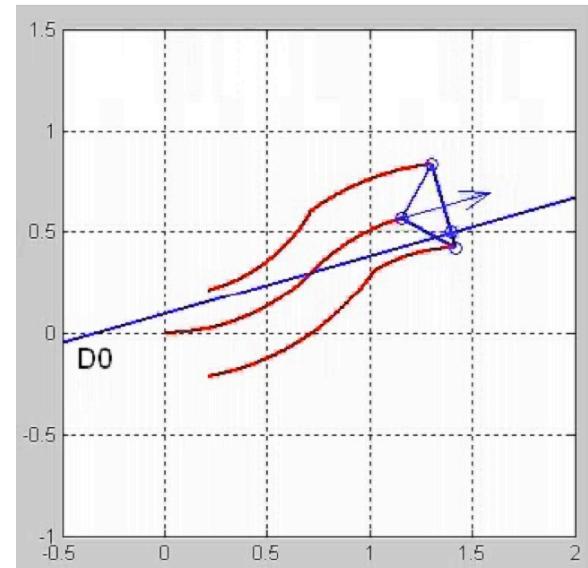
optimal solution to the EV problem : as before

## EV - example



$$l_1 = 0.3, \quad \phi_1 = \pi/4, \quad l_2 = 0.3, \quad \phi_2 = -\pi/4$$

obstacle in  $(1.4, 0.5)$



$C_a C_b$  type

## Family refinement

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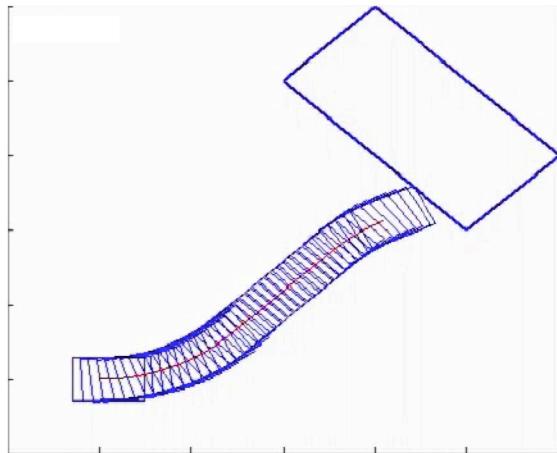
using continuity of the RS path parameters ( $a, b, e$ ) w.r.t. the robot parameter  $l_i$ , it is possible to show that

paths types (IV), (V), (VI), (VIII) are never optimal  
solutions of the problems VV, VE, EV

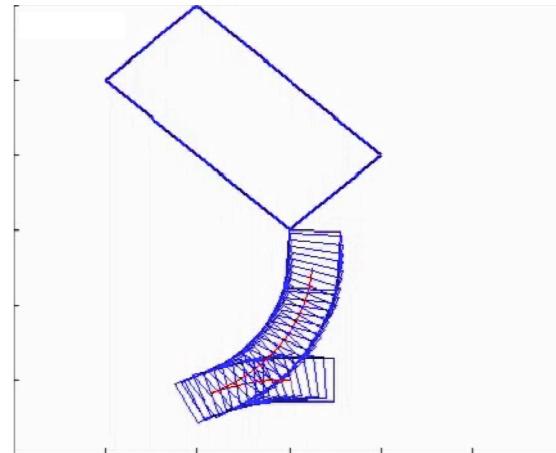
the search can be restricted to 26 path types

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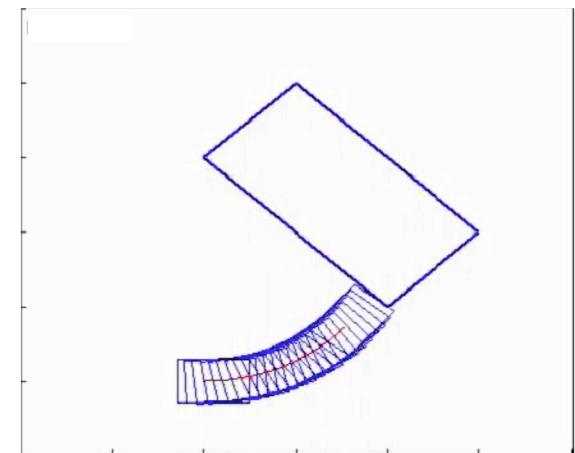
# examples



Vertex-Vertex



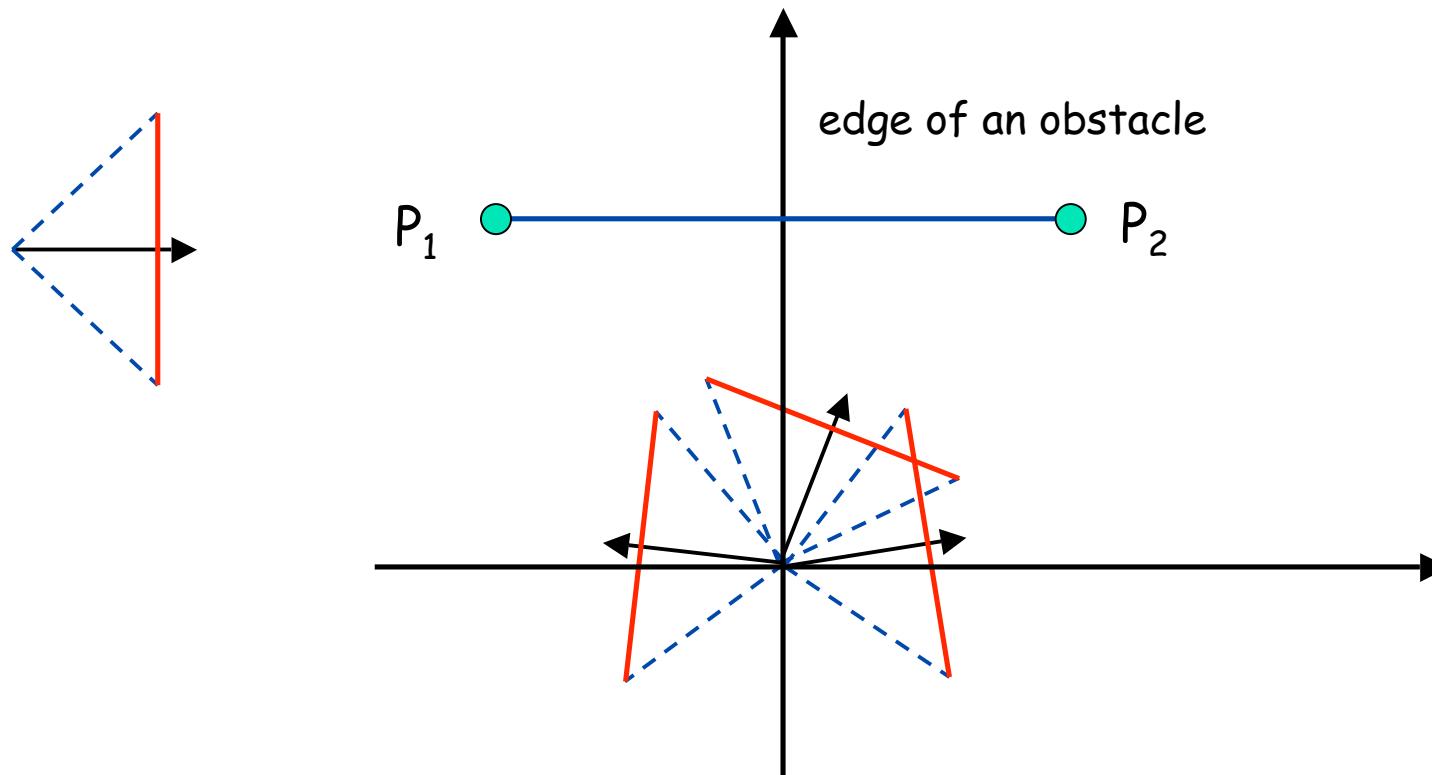
Vertex-Edge



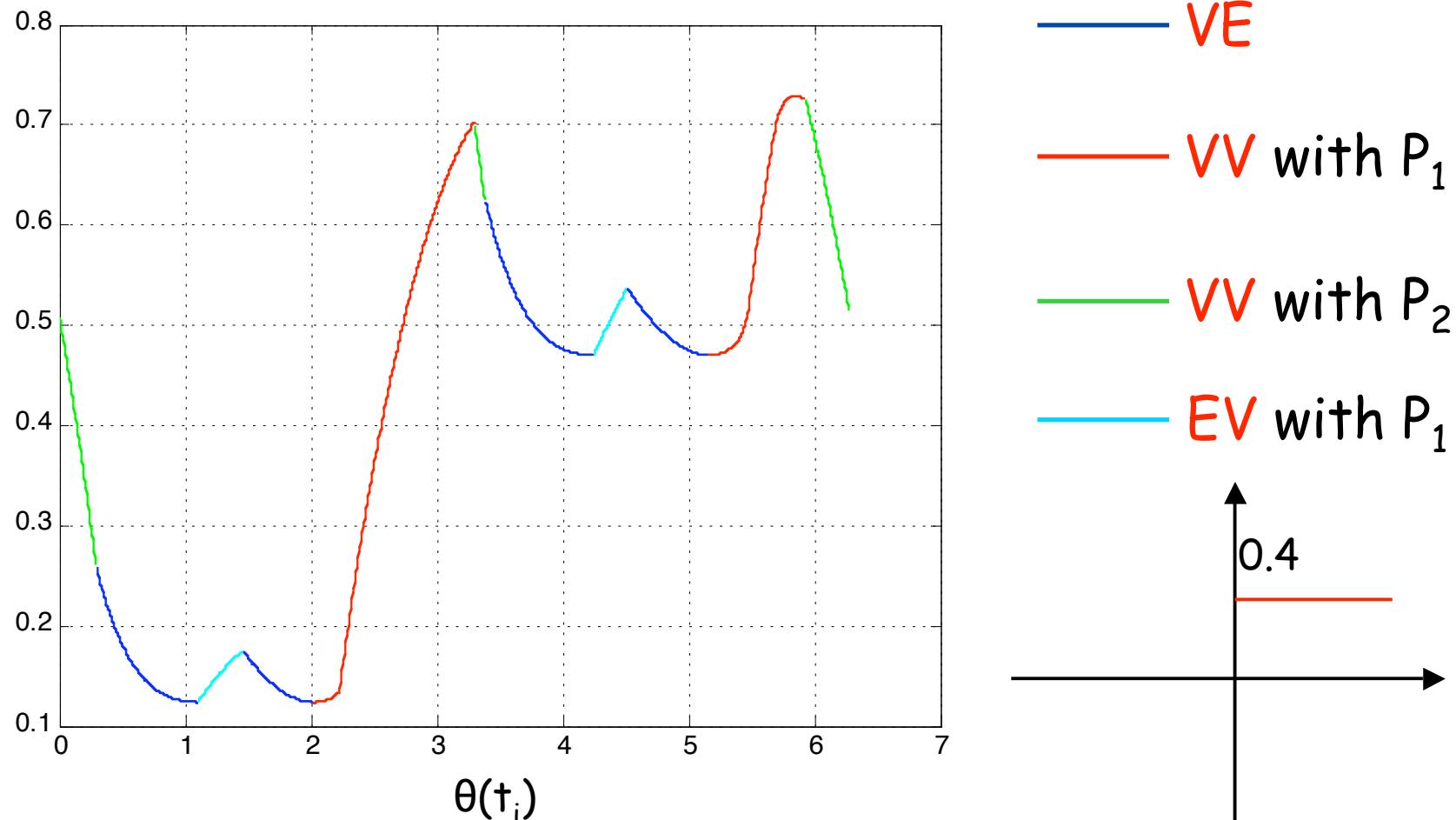
Edge-Vertex

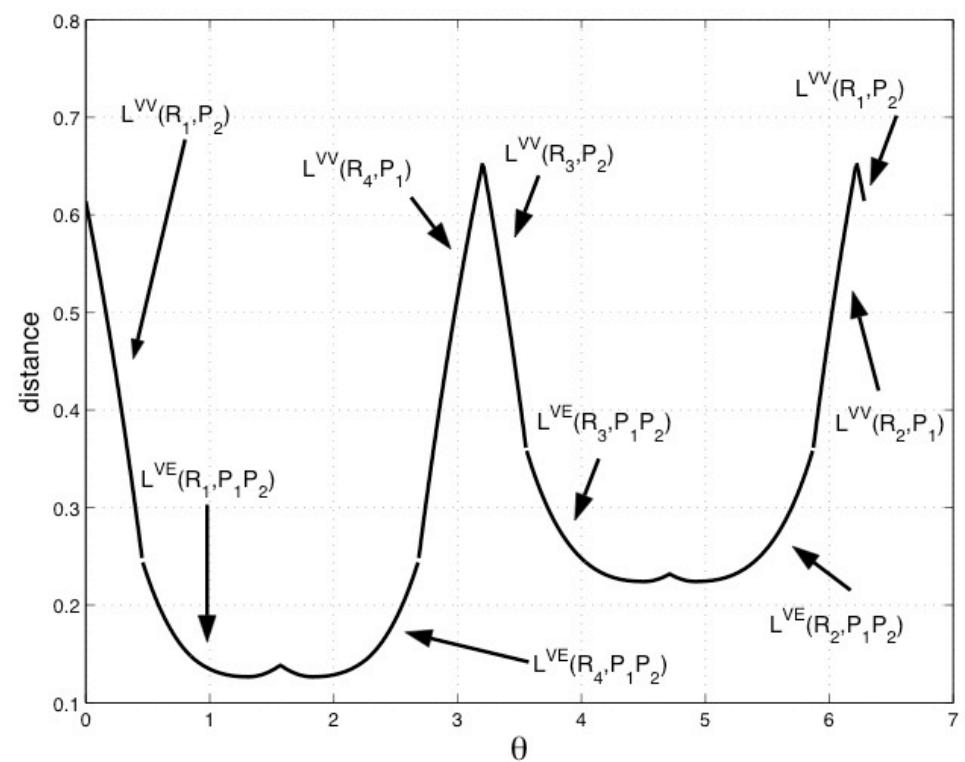
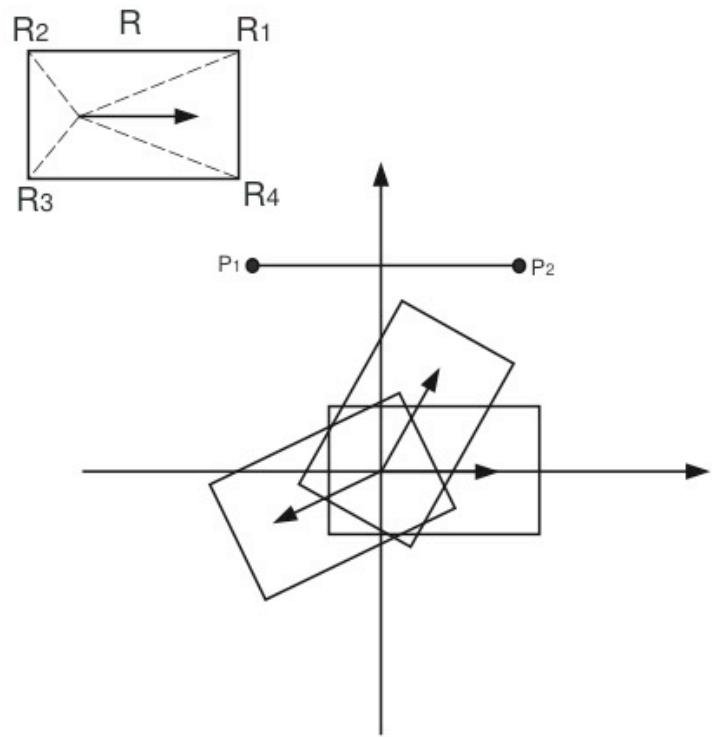
# smoothness

the defined distance function is piecewise smooth; the not derivable points are located at the switches between the  $L^{VV}$ ,  $L^{VE}$  and  $L^{EV}$  functions



$$P_1(0, 0.4) \quad P_2(0.3, 0.4)$$

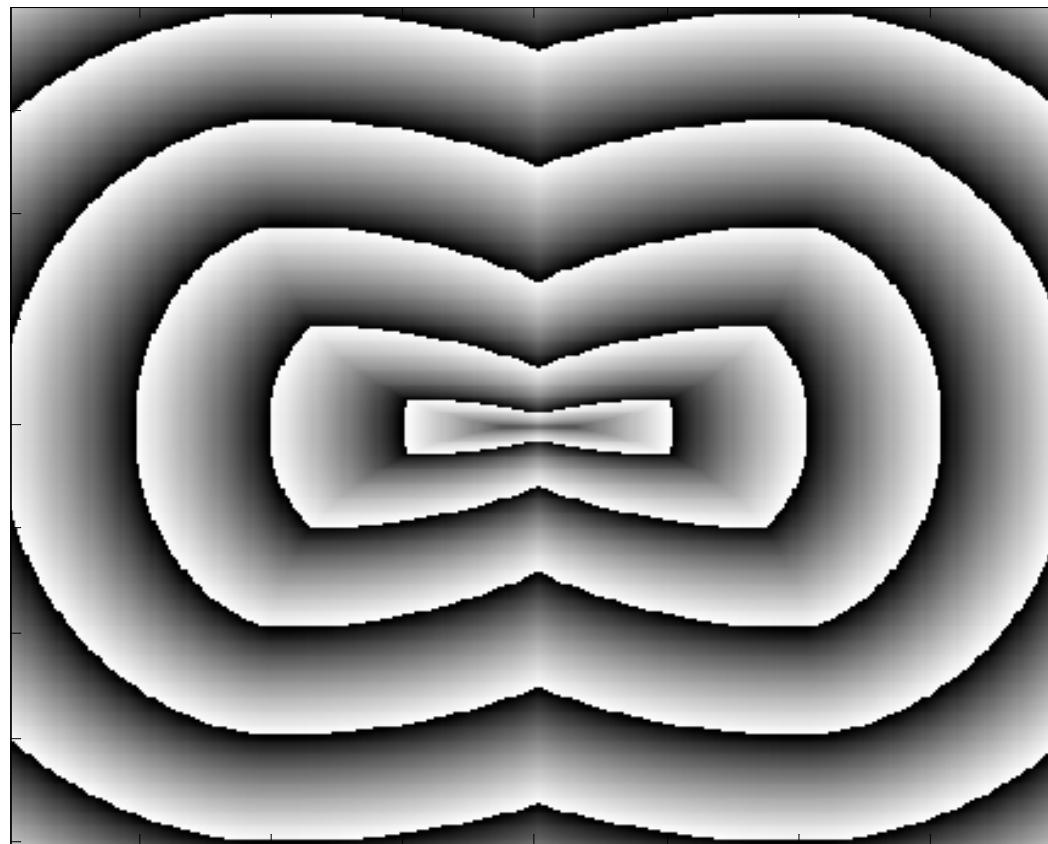


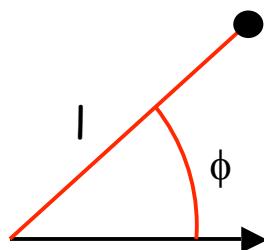


# level curves

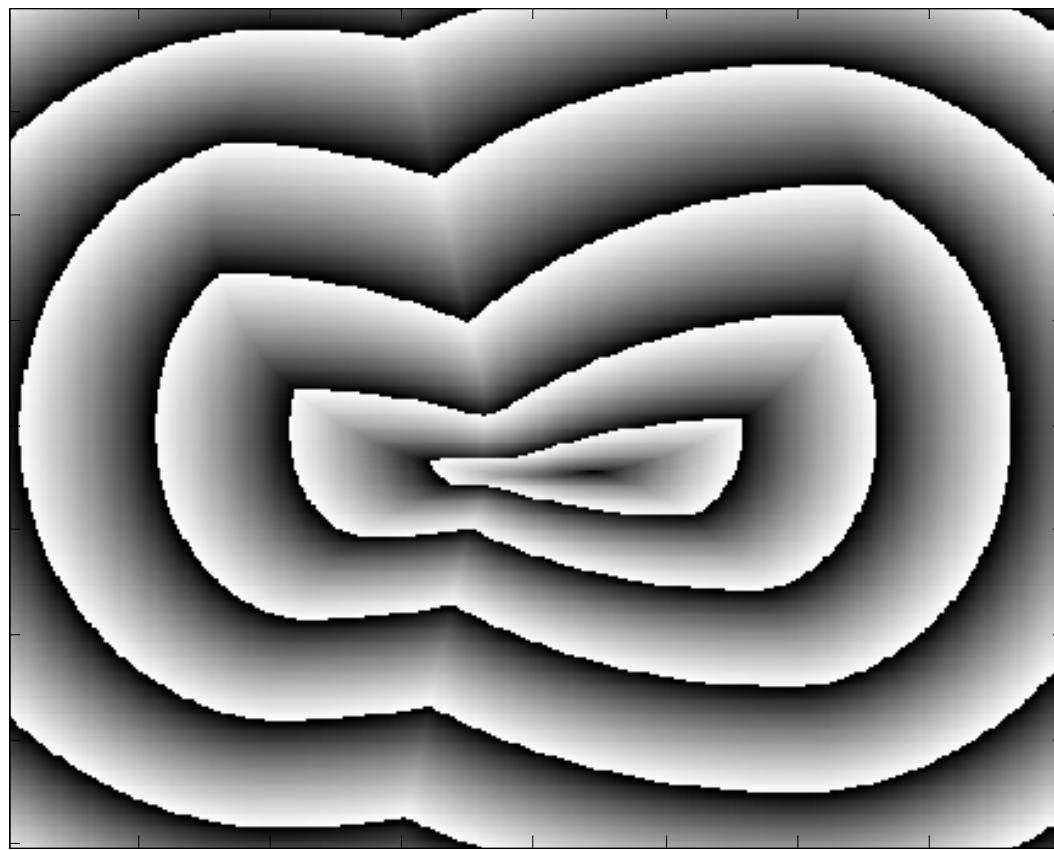
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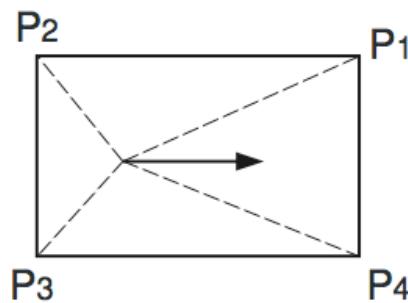
point robot:  $l=0, \phi=0$



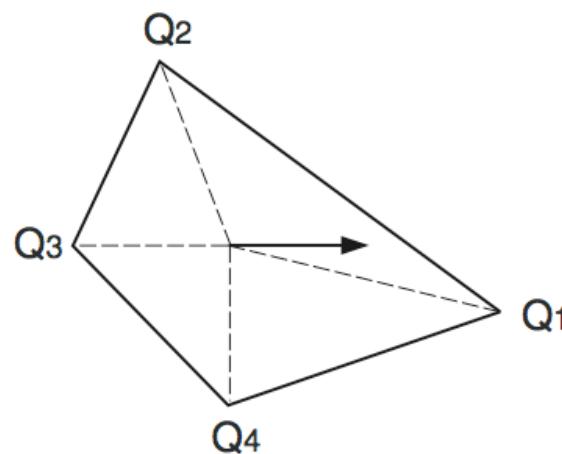
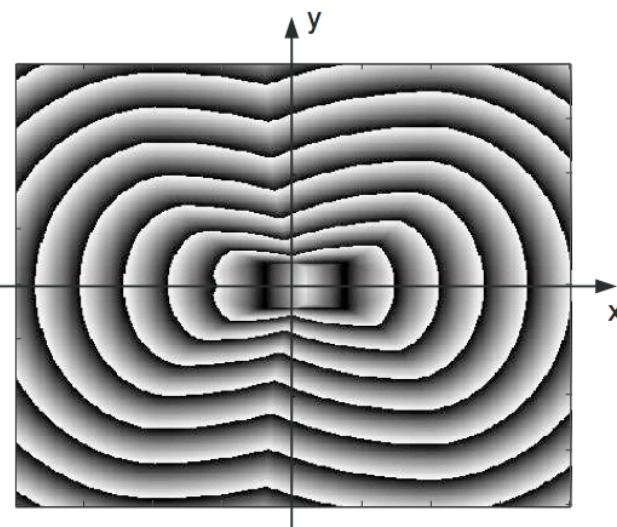


$$I=0.3, \phi = \pi/4$$

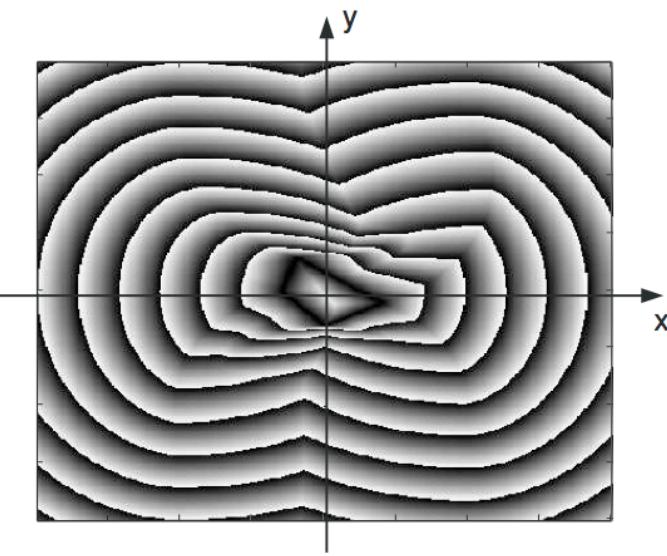




$R_1$



$R_2$



# Conclusion

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## main result

characterization of the shortest paths between  
polygonal RS car-like robot and polygonal obstacles

- analytic expression of the distance function
- reduction of the sufficient family of optimal paths

## future work

- extend to general manifolds in configuration space
- apply to motion planning