

Fast Computation of Robot-Obstacle Interactions in Nonholonomic Trajectory Deformation

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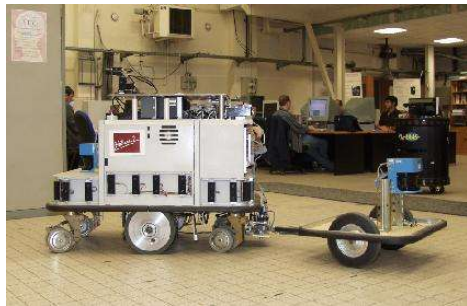
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Motivation and context

- Navigation in cluttered environment for nonholonomic systems

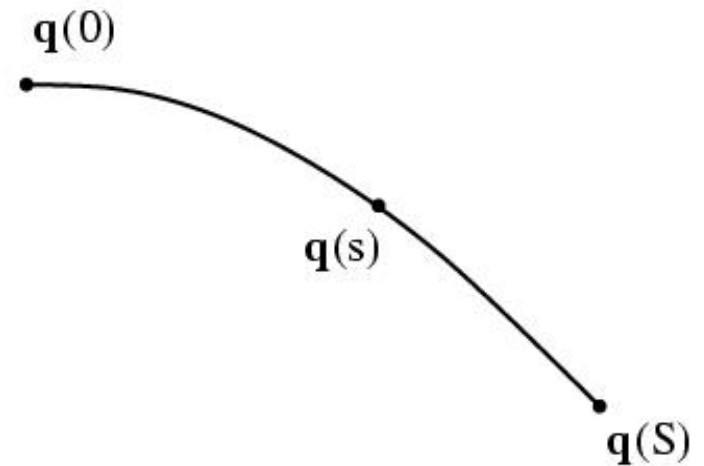
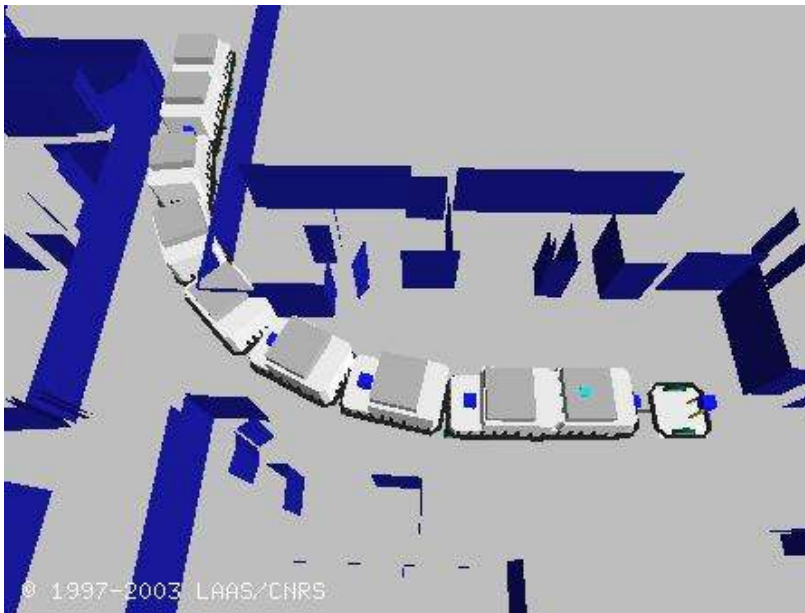


- Reactive obstacle avoidance
 - poor localization and map imprecision
 - unexpected obstacles

Nonholonomic trajectory deformation

Deform the trajectory in order to:

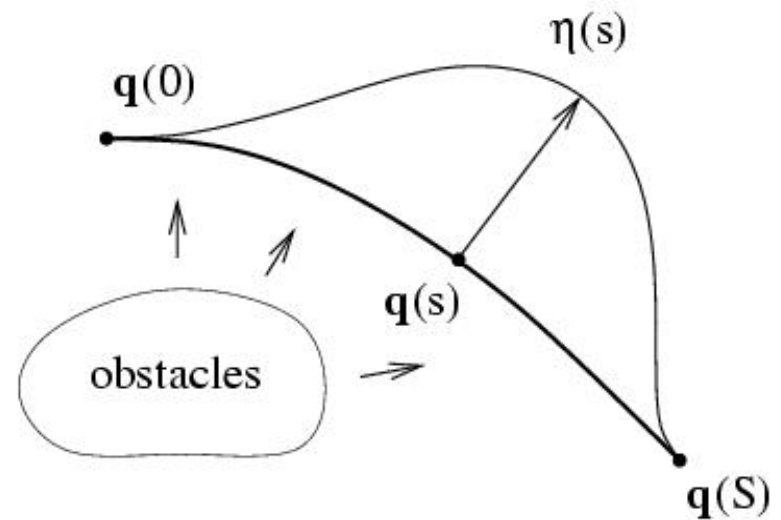
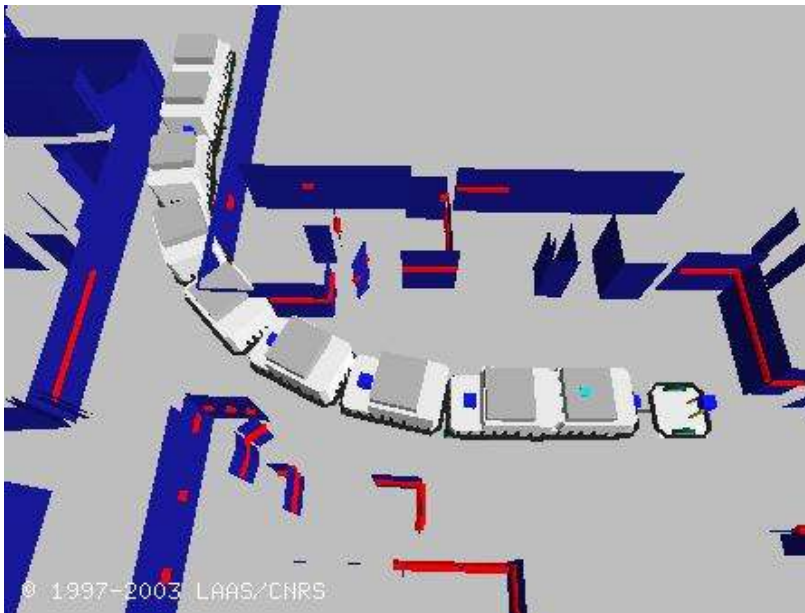
- move away from obstacles
- keep nonholonomic constraints satisfied



Nonholonomic trajectory deformation

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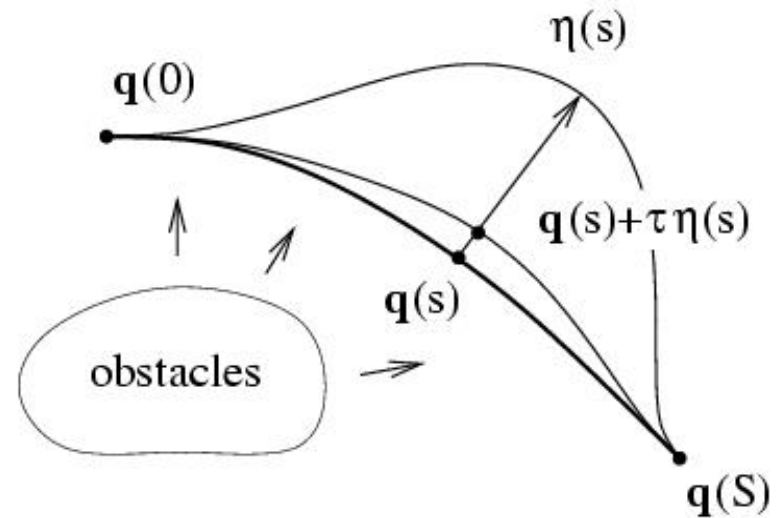
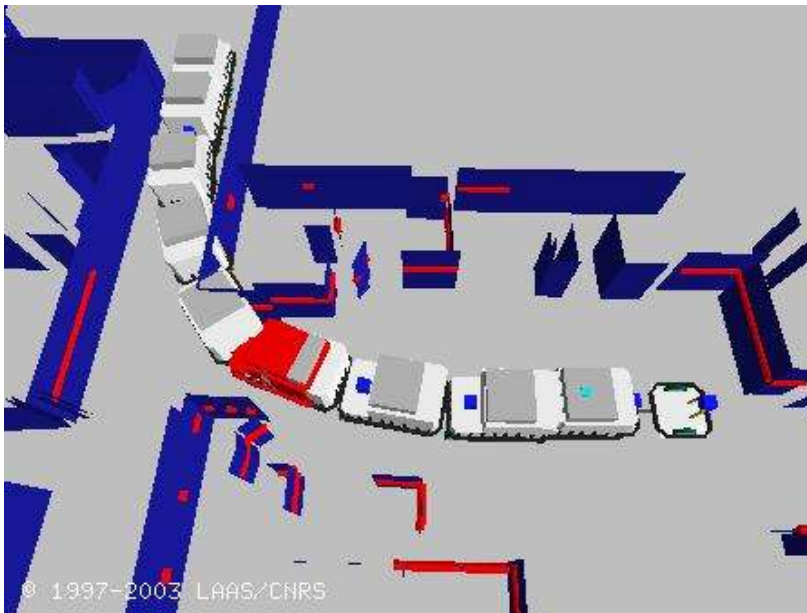
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Nonholonomic trajectory deformation

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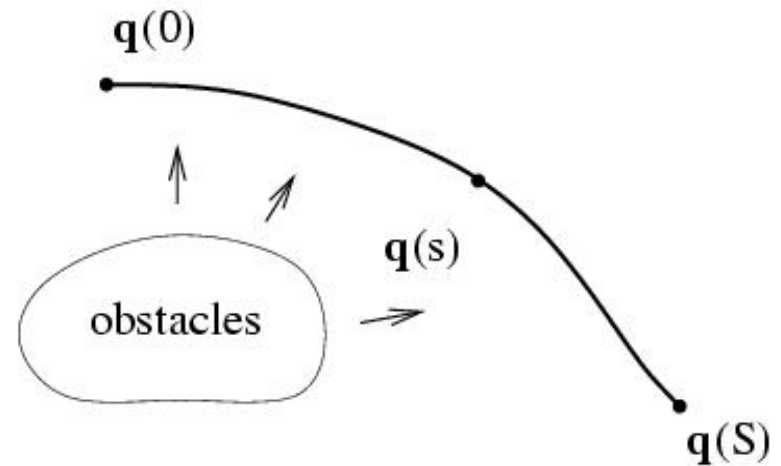
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Nonholonomic trajectory deformation

Deform the trajectory in order to:

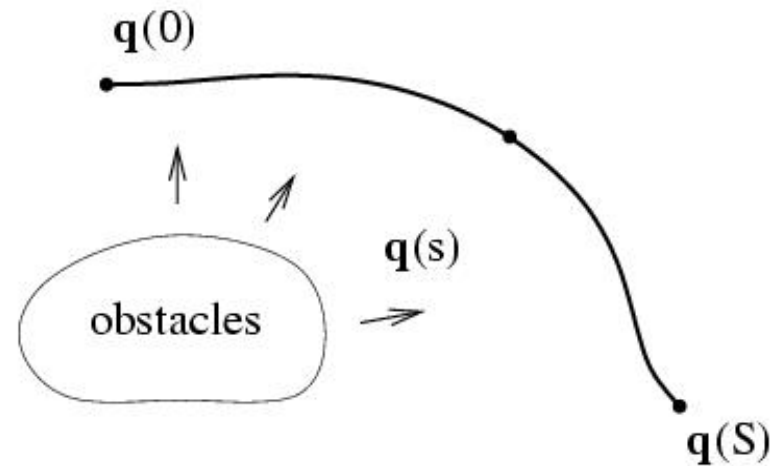
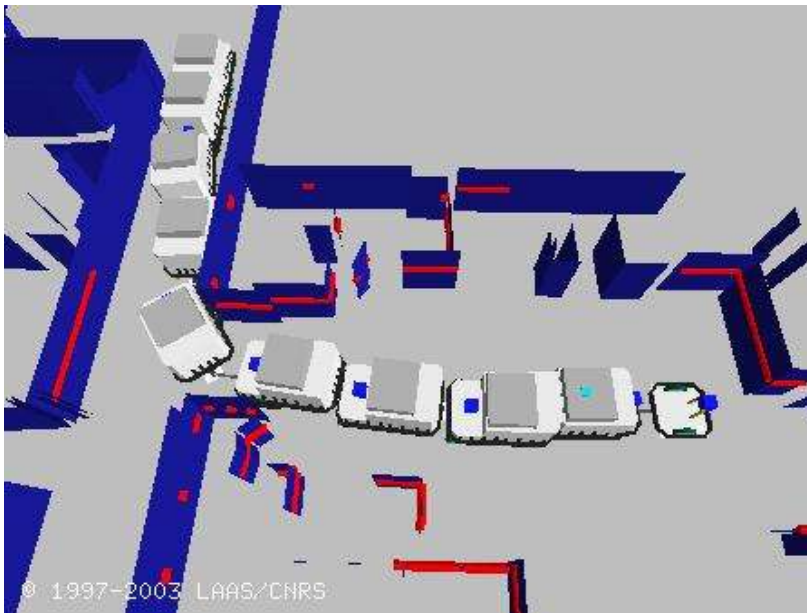
- move away from obstacles
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Nonholonomic trajectory deformation

Deform the trajectory in order to:

- move away from obstacles
- keep nonholonomic constraints satisfied



Direction of deformation

How to compute a direction of deformation that moves away from obstacles ?

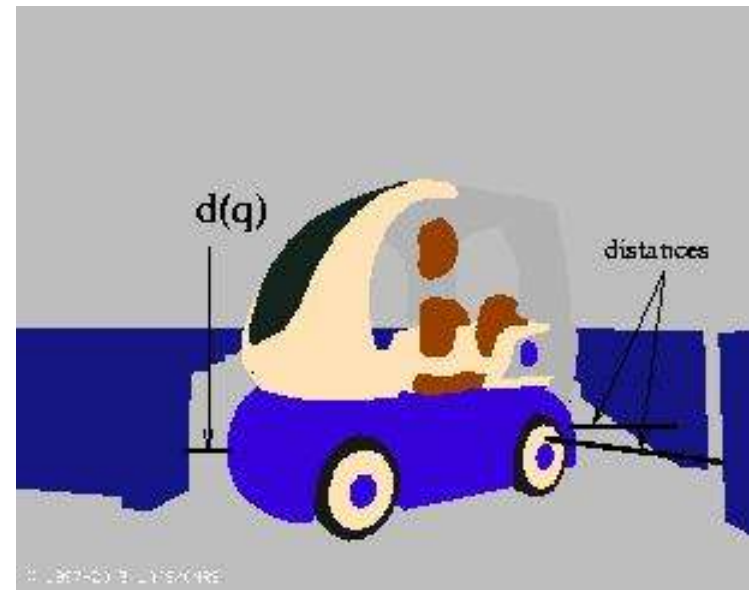
Direction of deformation

How to compute a direction of deformation that moves away from obstacles ?

Artificial potential fields over configuration space [Khatib 86]

$$U(\mathbf{q}(s)) = \frac{1}{d(\mathbf{q}(s))}$$

configurations close to obstacles
have a higher potential



Direction of deformation

- Potential of a trajectory

$$V(\mathbf{q}) = \int_0^S U(\mathbf{q}(s)) ds$$

move away from obstacles \Leftrightarrow make the potential decrease

Direction of deformation

- Potential of a trajectory

$$V(\mathbf{q}) = \int_0^S U(\mathbf{q}(s)) ds$$

move away from obstacles \Leftrightarrow make the potential decrease

- Gradient of the potential

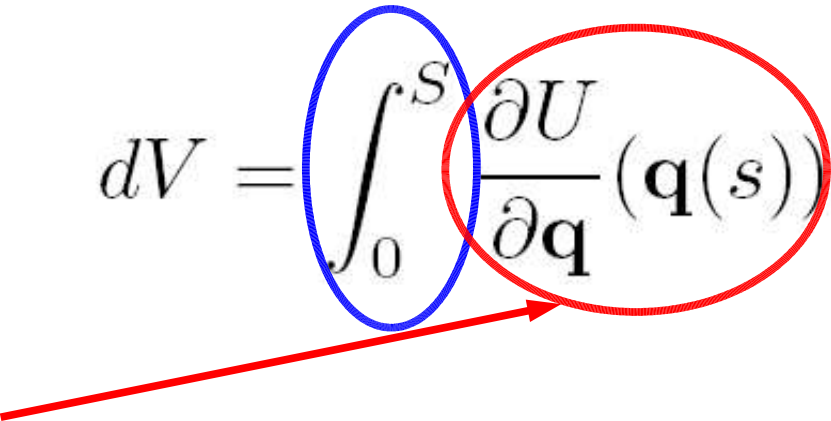
$$dV = \int_0^S \frac{\partial U}{\partial \mathbf{q}}(\mathbf{q}(s)) \eta(s) ds$$

compute $\eta(s)$ such that dV it is negative

Contribution

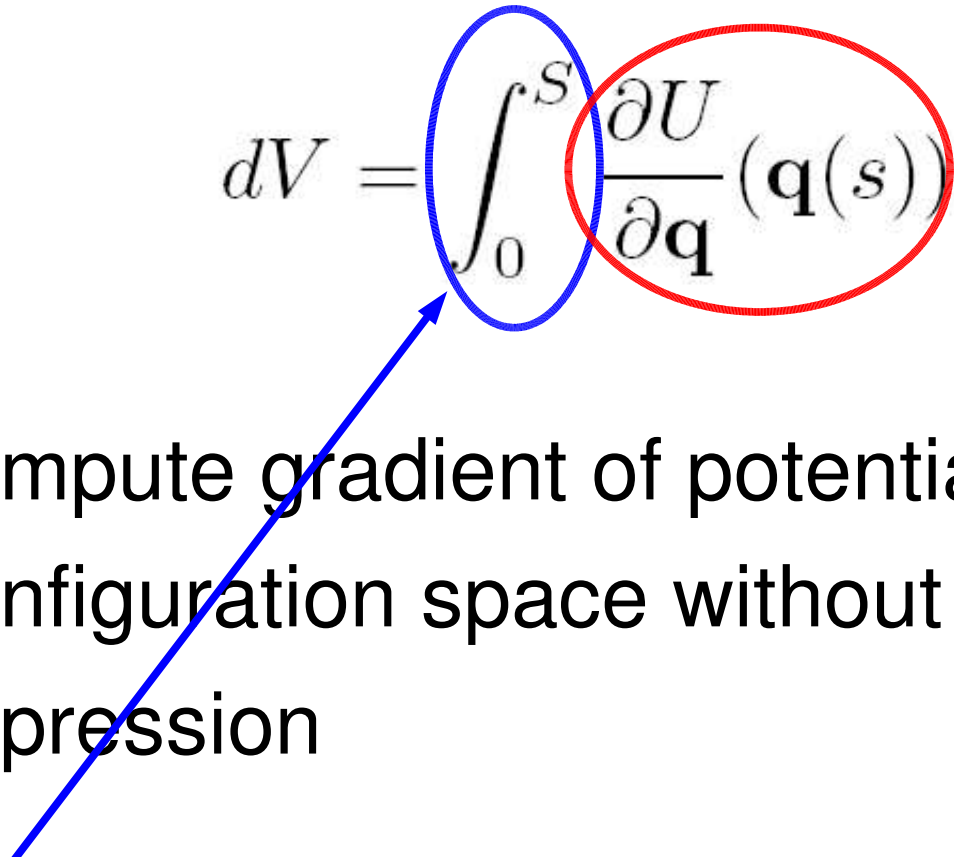
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Contribution

$$dV = \int_0^S \frac{\partial U}{\partial \mathbf{q}}(\mathbf{q}(s)) \eta(s) ds$$


- compute gradient of potential over configuration space without closed-form expression

Contribution

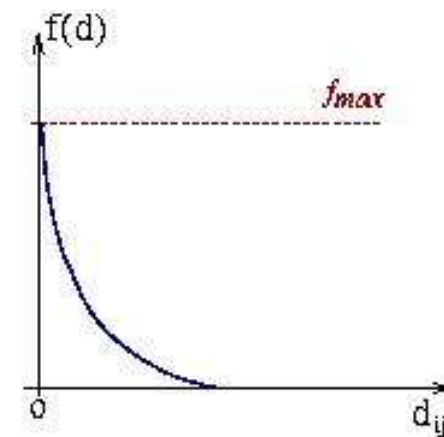
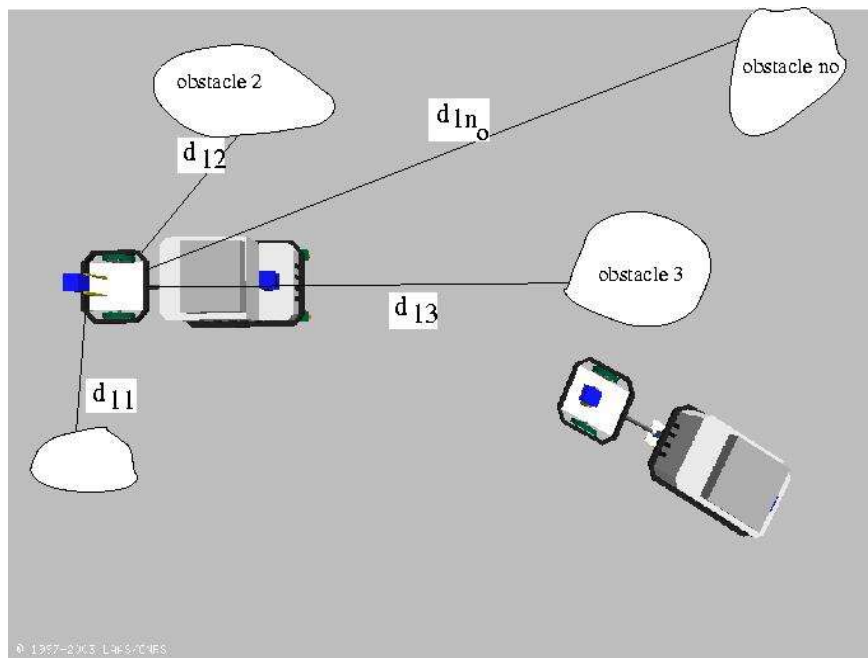
$$dV = \int_0^S \frac{\partial U}{\partial \mathbf{q}}(\mathbf{q}(s)) \eta(s) ds$$


- compute gradient of potential over configuration space without closed-form expression
- optimize computation over the trajectory using spacial coherence

Gradient of Potential

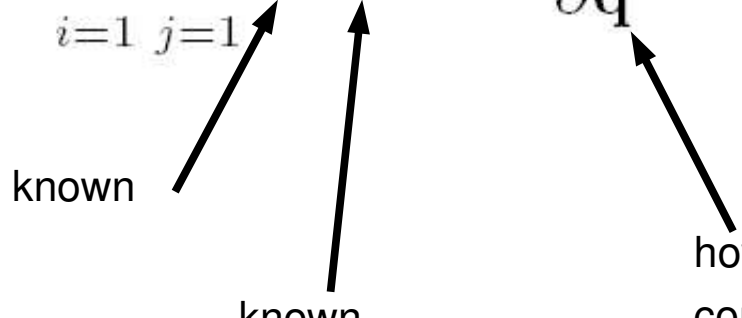
- $U(\mathbf{q})$ is a function of distance to obstacles:

$$U(\mathbf{q}) = \sum_{i=1}^{n_b} \sum_{j=1}^{n_o} f(d_{ij}(\mathbf{q}))$$



Gradient of Potential

- We need the gradient of the potential of a configuration

$$\frac{\partial U}{\partial \mathbf{q}}(\mathbf{q}) = \sum_{i=1}^{n_b} \sum_{j=1}^{n_o} f'(d_{ij}(\mathbf{q})) \frac{\partial d_{ij}}{\partial \mathbf{q}}(\mathbf{q})$$


known

known

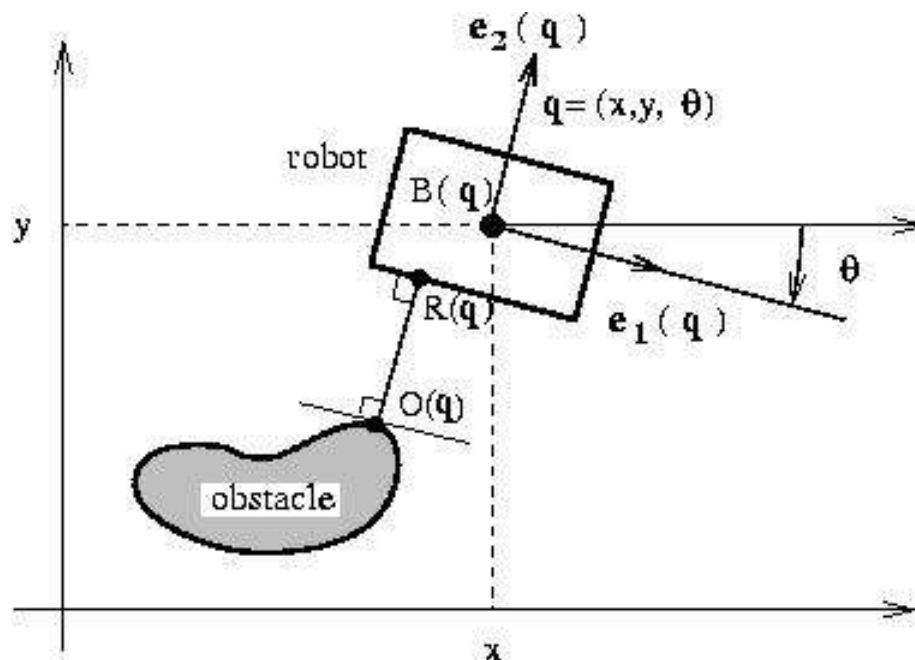
how to compute it ?

The diagram shows three arrows pointing from text labels to specific parts of the equation. The first arrow, labeled 'known', points to the summation indices $i=1$ and $j=1$. The second arrow, also labeled 'known', points to the function $f'(d_{ij}(\mathbf{q}))$. The third arrow, labeled 'how to compute it ?', points to the partial derivative term $\frac{\partial d_{ij}}{\partial \mathbf{q}}(\mathbf{q})$.

Gradient of Potential

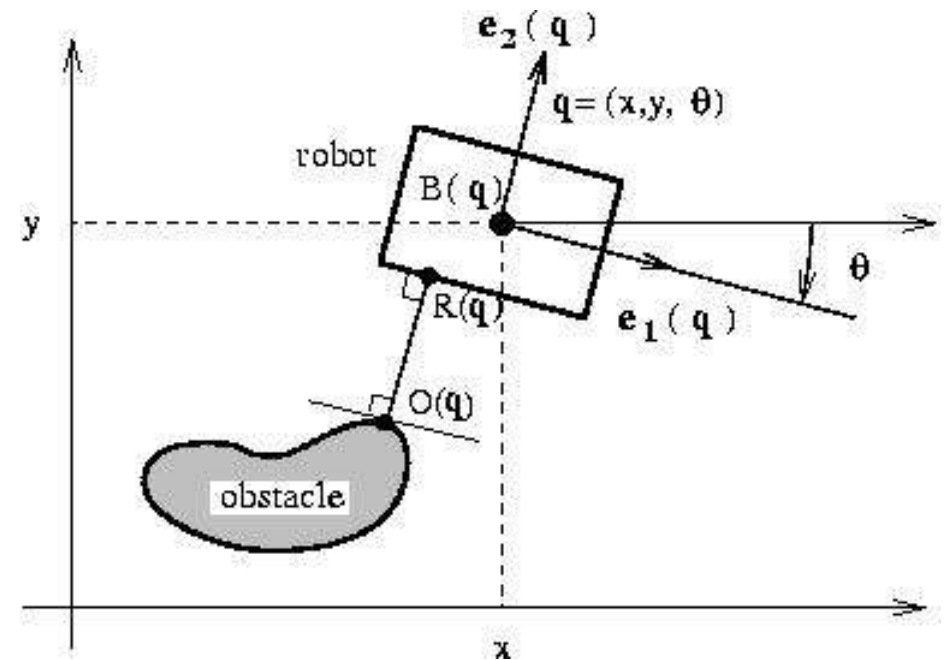
- derivative of distance:

$$\frac{\partial d_{ij}}{\partial \mathbf{q}}(\mathbf{q}) = \frac{(\mathbf{O}(\mathbf{q}) - \mathbf{R}(\mathbf{q}))^T}{\|\mathbf{O}(\mathbf{q}) - \mathbf{R}(\mathbf{q})\|} \left(\frac{\partial \mathbf{O}}{\partial \mathbf{q}}(\mathbf{q}) - \frac{\partial \mathbf{R}}{\partial \mathbf{q}}(\mathbf{q}) \right)$$



Composition of motions

$$\mathbf{R}(\mathbf{q}) = \sum_{l=1}^d \rho_l(\mathbf{q}) \mathbf{e}_l(\mathbf{q})$$



$$\frac{\partial \mathbf{R}}{\partial \mathbf{q}}(\mathbf{q}) = \sum_{l=1}^d \frac{\partial \rho_l}{\partial \mathbf{q}}(\mathbf{q}) \mathbf{e}_l(\mathbf{q}) + \sum_{l=1}^d \rho_l(\mathbf{q}) \frac{\partial \mathbf{e}_l(\mathbf{q})}{\partial \mathbf{q}}$$

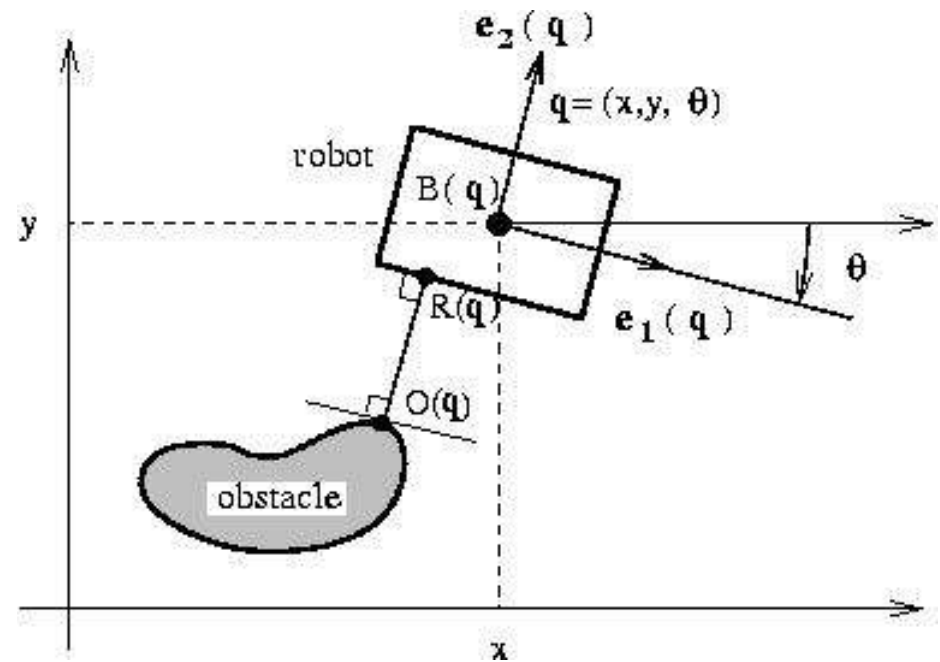
← motion of $R(\mathbf{q})$ on the robot

← absolute motion of coinciding point

Composition of motions

- motion of $R(q)$ is orthogonal to

$$(\mathbf{O}(q) - \mathbf{R}(q))^T$$



$$\frac{\partial \mathbf{R}}{\partial \mathbf{q}}(\mathbf{q}) = \sum_{l=1}^d \frac{\partial \rho_l}{\partial \mathbf{q}}(\mathbf{q}) \mathbf{e}_l(\mathbf{q}) +$$

← motion of $R(q)$ on the robot

$$\sum_{l=1}^d \rho_l(\mathbf{q}) \frac{\partial \mathbf{e}_l(\mathbf{q})}{\partial \mathbf{q}}$$

← absolute motion of coinciding point

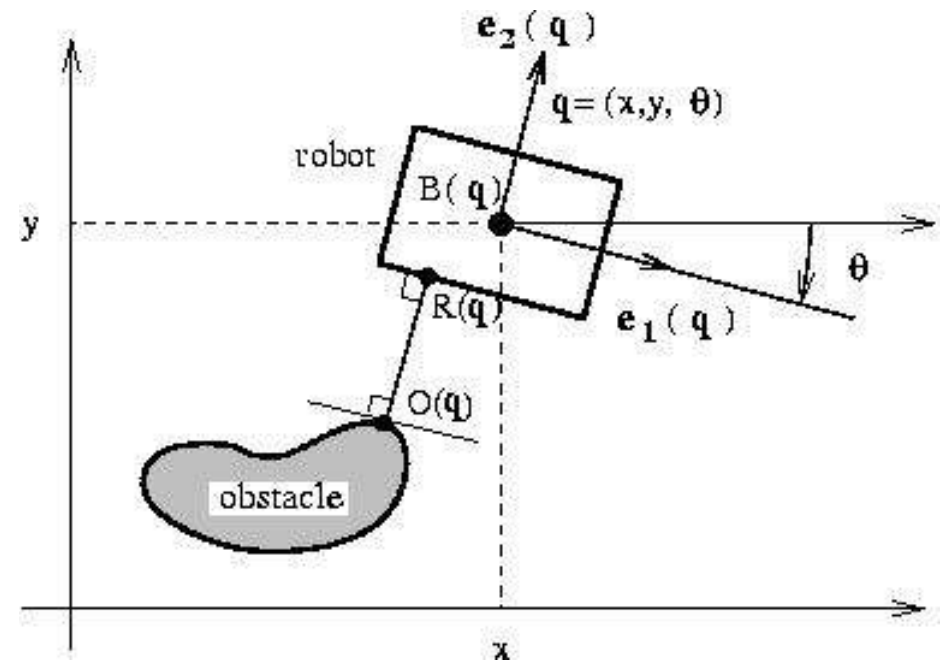
Composition of motions

- motion of $R(q)$ is orthogonal to:

$$(\mathbf{O}(q) - \mathbf{R}(q))^T$$

and we need to compute:

$$\frac{\partial d_{ij}}{\partial \mathbf{q}}(q) = \frac{(\mathbf{O}(q) - \mathbf{R}(q))^T}{\|\mathbf{O}(q) - \mathbf{R}(q)\|} \left(\frac{\partial \mathbf{O}}{\partial \mathbf{q}}(q) - \frac{\partial \mathbf{R}}{\partial \mathbf{q}}(q) \right)$$



$$\frac{\partial \mathbf{R}}{\partial \mathbf{q}}(q) = \sum_{l=1}^d \frac{\partial \rho_l}{\partial \mathbf{q}}(q) \mathbf{e}_l(q) +$$

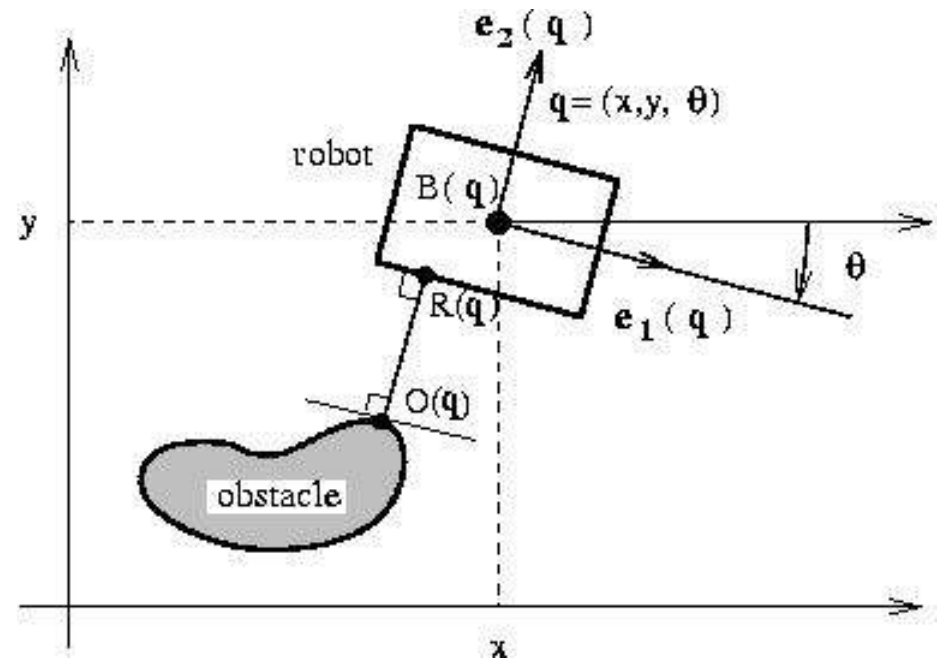
← motion of $R(q)$ on the robot

$$\sum_{l=1}^d \rho_l(q) \frac{\partial \mathbf{e}_l(q)}{\partial \mathbf{q}}$$

← absolute motion of coinciding point

Composition of motions

- same reasoning with obstacles
- fixed obstacles



$$\frac{\partial \mathbf{O}}{\partial \mathbf{q}}(\mathbf{q}) = \sum_{l=1}^d \frac{\partial \rho_l}{\partial \mathbf{q}}(\mathbf{q}) \mathbf{e}_l(\mathbf{q}) + \sum_{l=1}^d \rho_l(\mathbf{q}) \frac{\partial \mathbf{e}_l(\mathbf{q})}{\partial \mathbf{q}}$$

← ~~motion of $\mathbf{O}(\mathbf{q})$ on the obstacle~~

← ~~absolute motion of coinciding point~~

Gradient of configuration space potential field

- No closed-form required if it is a function of distance to obstacles
- compute $\frac{\partial \mathbf{R}_{\in body}}{\partial \mathbf{q}}$ only

Filter useless Robot-Obstacles Interaction pairs

$$dV = \left(\int_0^S \right) \frac{\partial U}{\partial \mathbf{q}}(\mathbf{q}(s)) \eta(s) ds$$

Filter useless Robot-Obstacles Interaction pairs

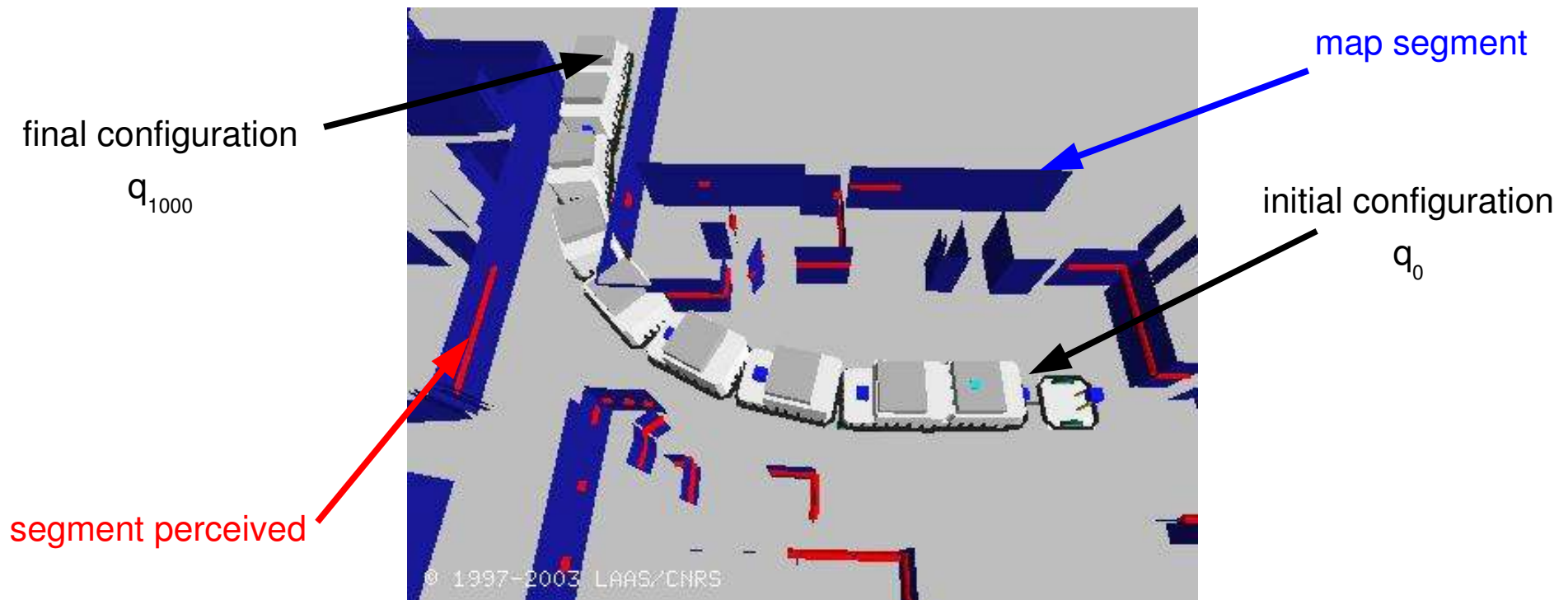
$$dV = \int_0^S \frac{\partial U}{\partial \mathbf{q}}(\mathbf{q}(s)) \eta(s) ds$$

Computations are done for all configurations
along the planned trajectory

How to reduce the complexity ?

Filter useless Robot-Obstacles Interaction pairs

- Given a sensor perception and a discretized trajectory

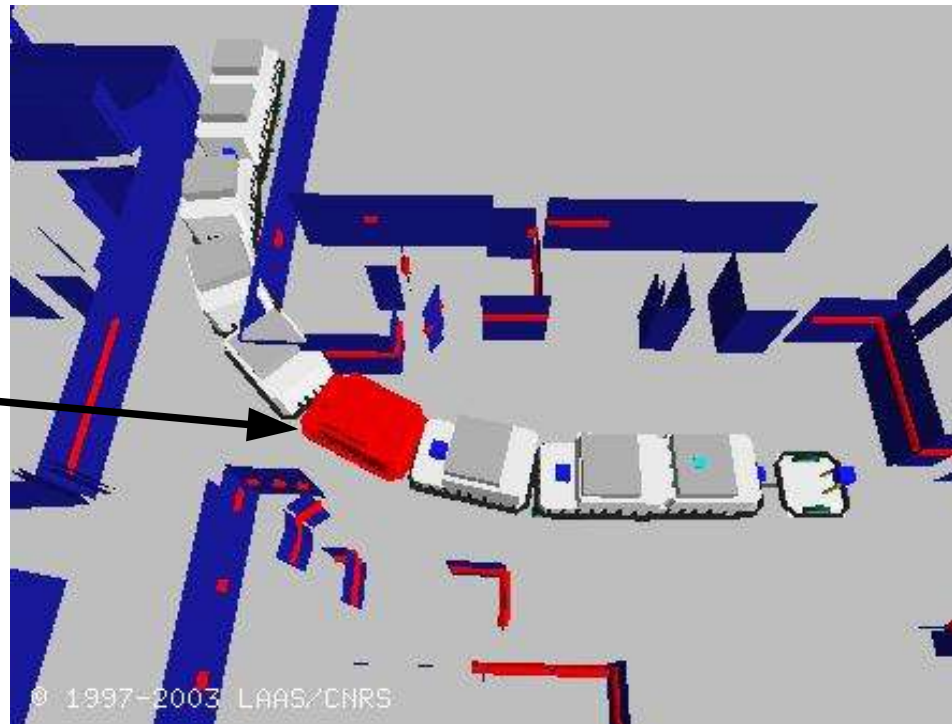


Filter useless Robot-Obstacles Interaction pairs

1. compute index of collision ($n_{conf} * n_{obstacles}$)

each configuration is tested against collision with each obstacle

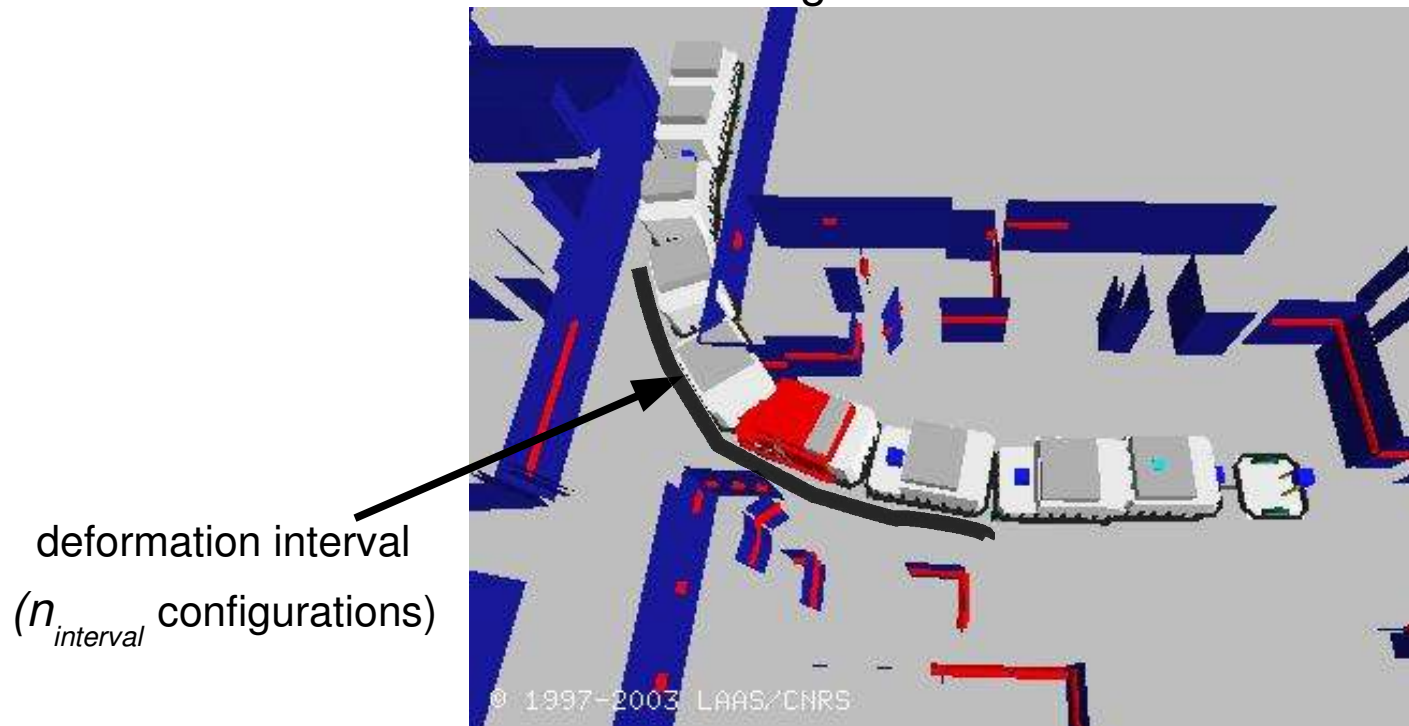
first configuration
in collision



Filter useless Robot-Obstacles Interaction pairs

2. compute the gradient of the potential field $(n_{interval} * n_{obstacles})$

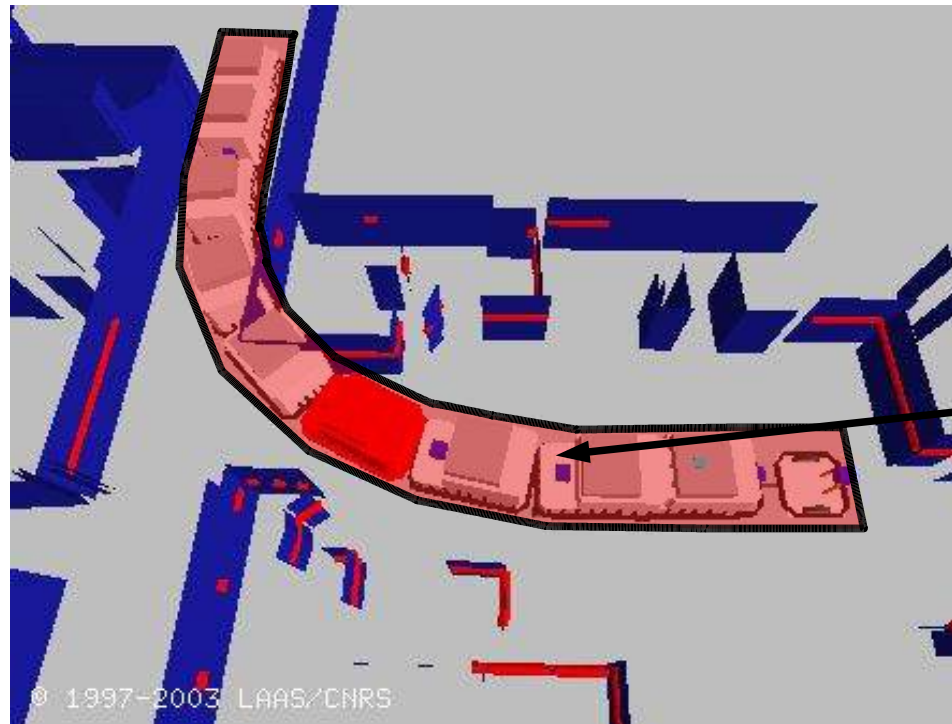
gradient of potential created by every obstacle is computed for each configuration
along the deformation interval



Useless Robot-Obstacle pairs

- Most of the robot-obstacle pairs are useless:

collision : minimal clearance distance ρ



collision tunnel for body trailer
swept volume = $w_{\text{trailer}} + 2\rho$

Useless Robot-Obstacle pairs

- Most of the robot-obstacle pairs are useless:

gradient of the potential field : maximal distance of influence ρ



influence distance tunnel
swept volume = $w_{\text{trailer}} + 2\rho$

Filter using spatial coherence

- spatial coherence:
 - obstacles “far” from q_s are still “far” from q_{s+1}

maximal distance traveled by points of body B_i between q_s and q_{s+1} : Δ

distance between 2 compact subsets : $d(\mathcal{A}, \mathcal{B}) = \min_{a \in \mathcal{A}, b \in \mathcal{B}} \|b - a\|$

Filter using spatial coherence

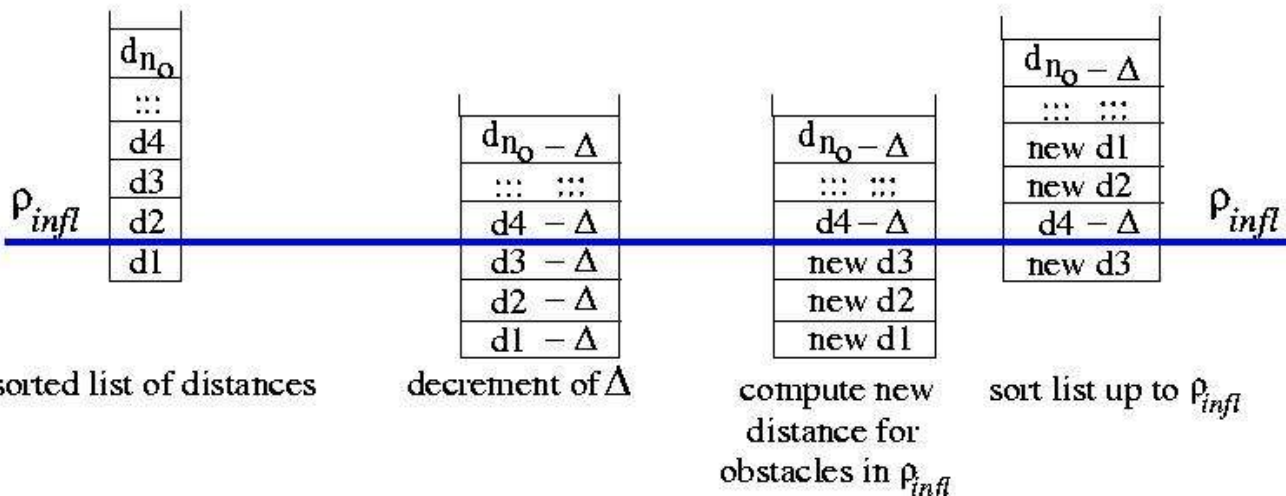
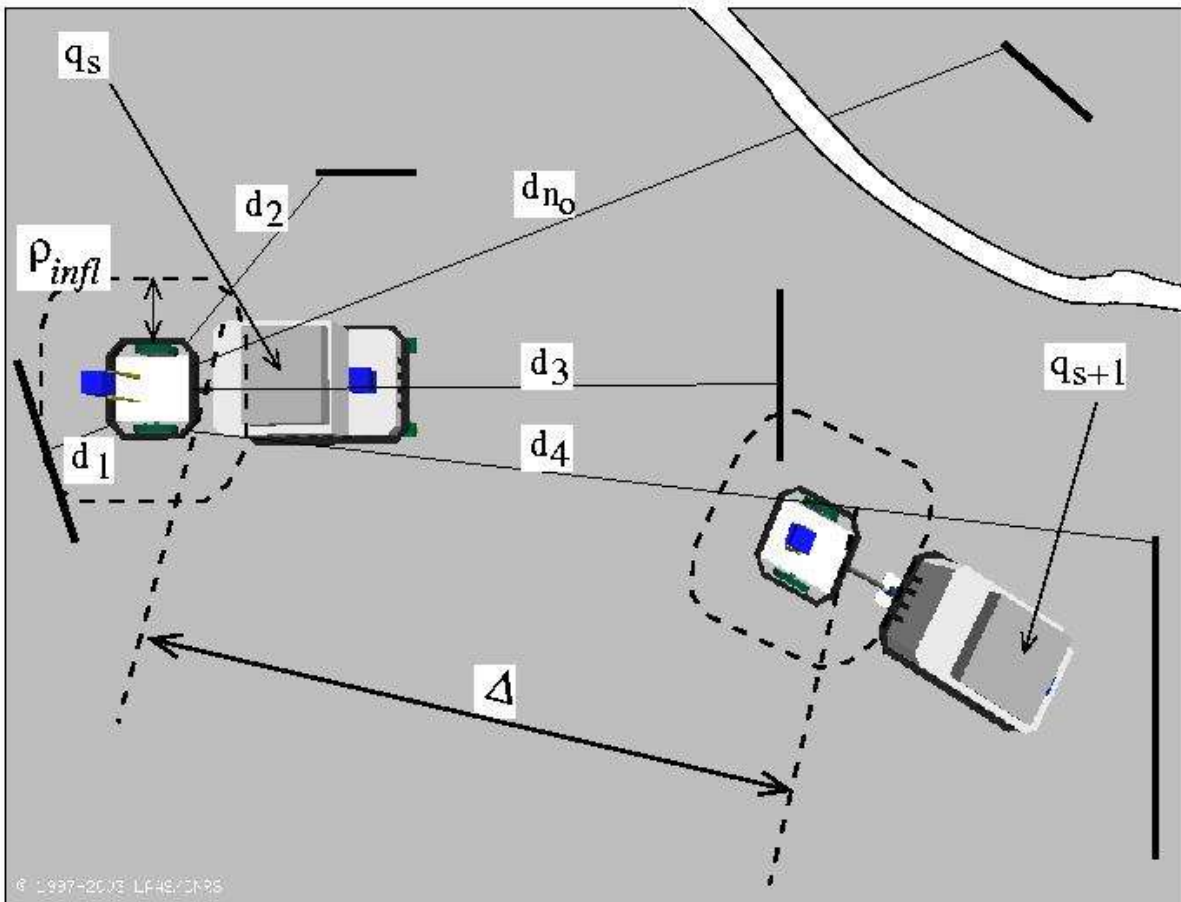
- **spatial coherence:** example with the trailer

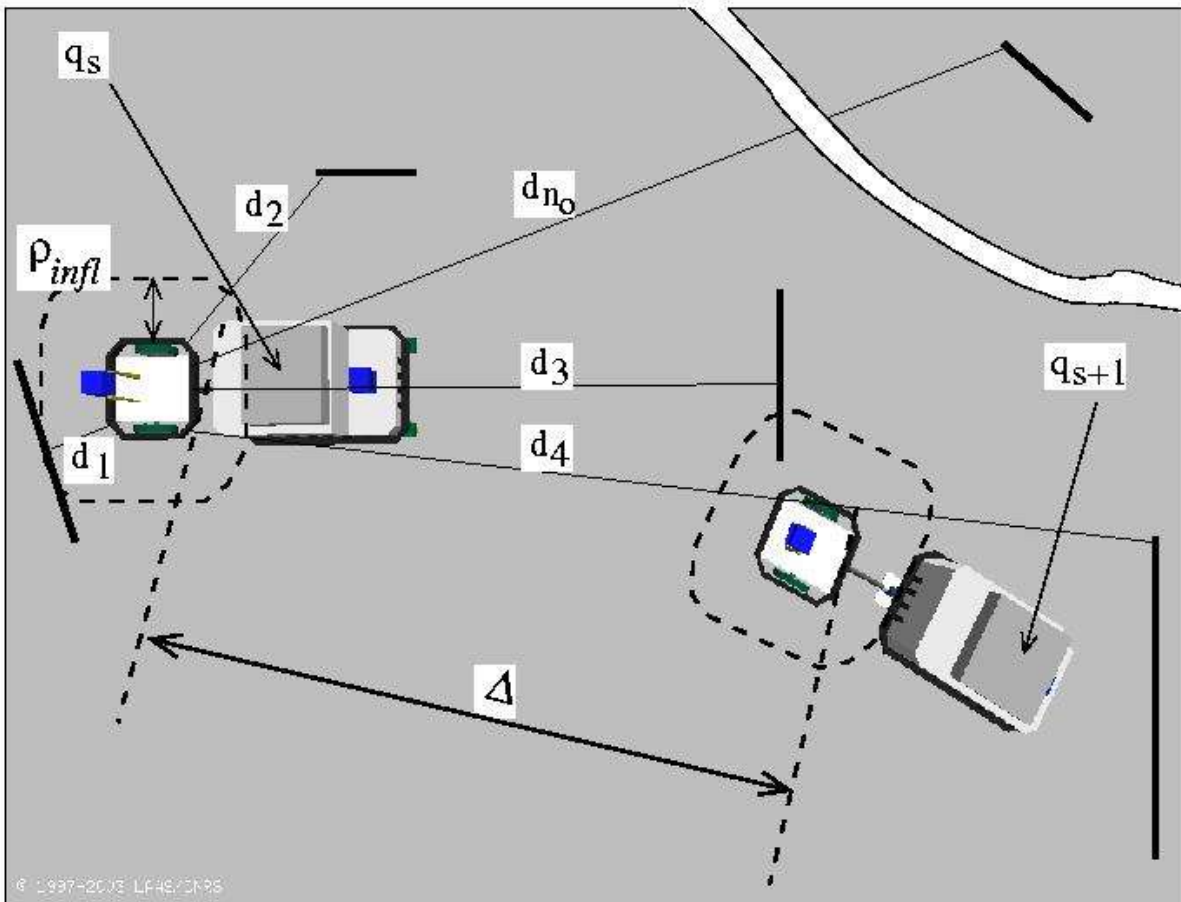
$$d(\mathcal{B}_i(\mathbf{q}_1), \mathcal{O}) \geq d_1 \quad \Rightarrow \quad d(\mathcal{B}_i(\mathbf{q}_2), \mathcal{O}) \geq \max(d_1 - \Delta, 0)$$



- compute a sorted list of distances to obstacles

distance $< \rho_{infl}$ \Rightarrow usefull robot-obstacle pair





- compute a sorted list of distances to obstacles

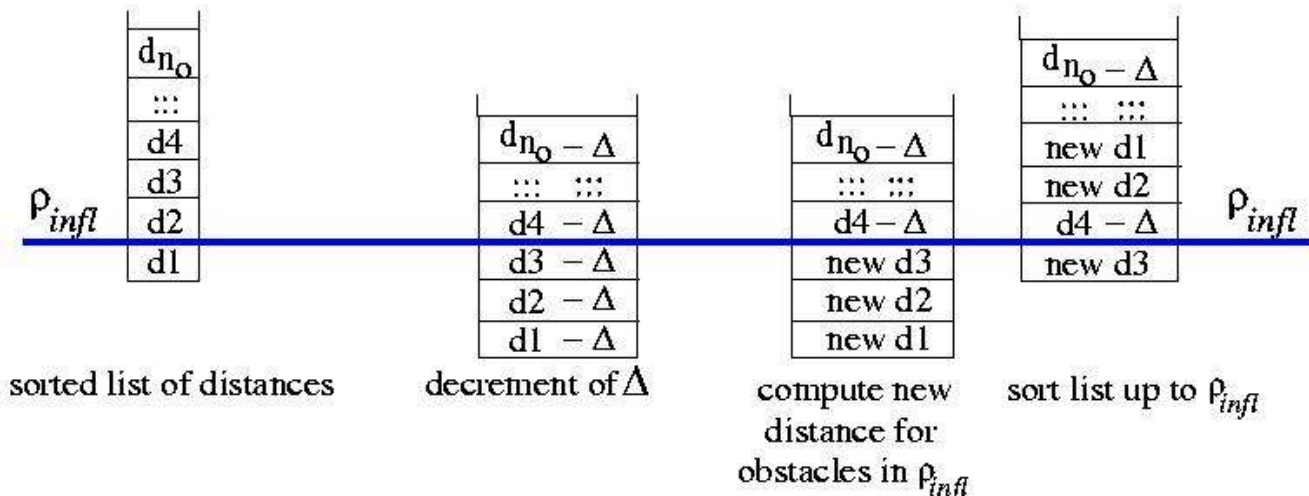
distance $< \rho_{infl}$ \Rightarrow usefull robot-obstacle pair

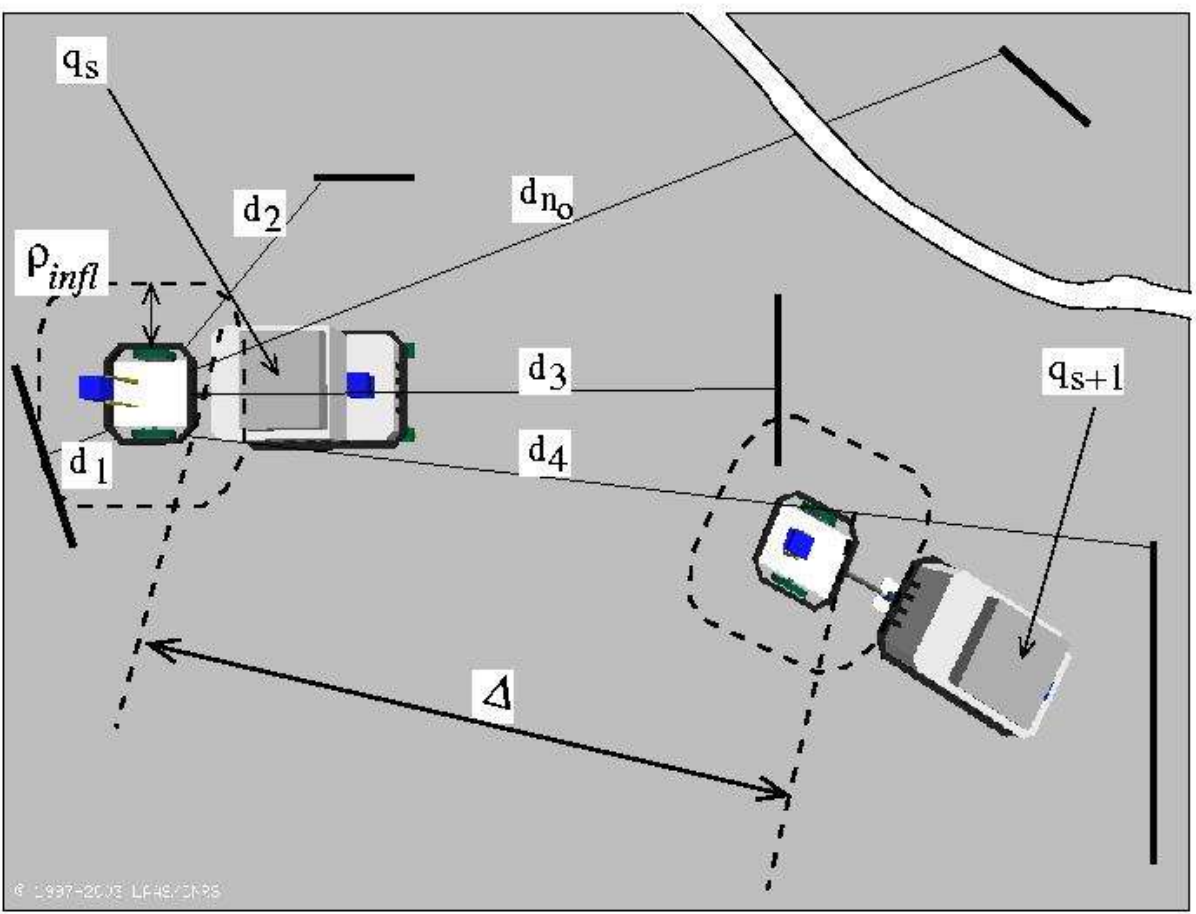
- between two configurations

subtract Δ to all elements

distance $< \rho$ \Rightarrow recompute exact distance

insert it in the list





- compute a sorted list of distances to obstacles

distance < ρ_{infl} => usefull robot-obstacle pair

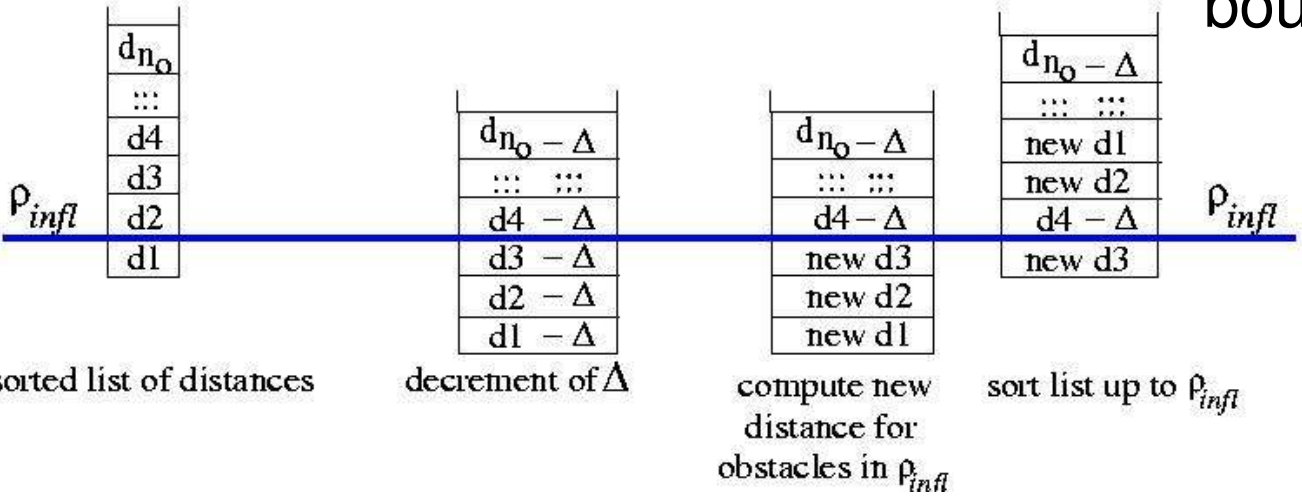
- between two configurations

subtract Δ to all elements

distance < ρ => recompute exact distance

insert it in the list

- manage a sorted list of lower bounds



Experimental Results

- Implementation on several robots:
 - hilare2 (with trailer), cycab (car-like), dala (rover)
- Time of computation gain:
 - collision checking: divided by 10
 - gradient of potential: divided by 6

Conclusion

- Gradient of potential in configuration space:
 - no closed-form expression if the potential is a function of the distance to obstacles
- Filter useless robot-obstacle interactions pairs:
 - maximum influence distance of interactions
 - spatial coherence
- Future work: complexity of the filtering algorithm