

# Motion and path planning in a nutshell

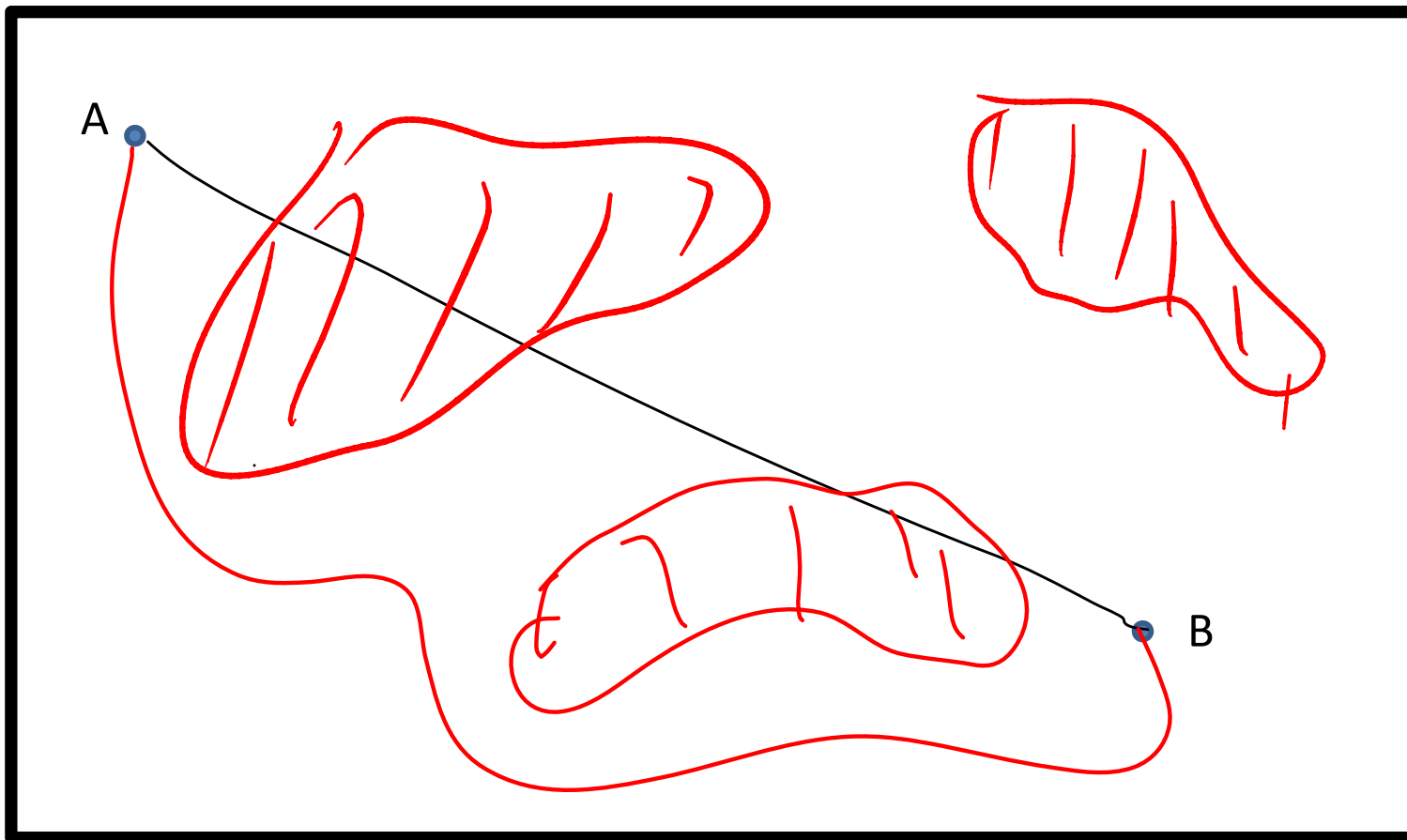
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MAE

Guest lecture: CS 4758/6758

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easy



not as easy

# “How do I get to point B?”

- Motion planning
  - Bug algorithms
  - Roadmaps, cell decomposition
  - Potential functions
  - Sampling-based methods

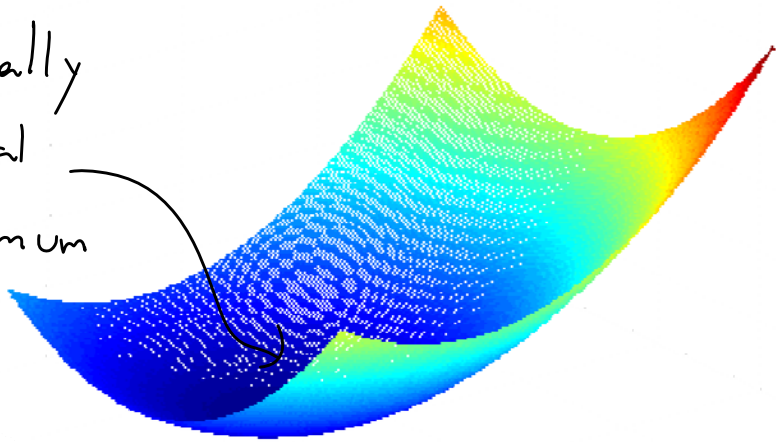
# “How do I get to point B?”

- Motion planning
  - Bug algorithms
  - Roadmaps, cell decomposition
  - **Potential functions (“vanilla” potential functions)**
  - **Sampling-based methods (RRT)**

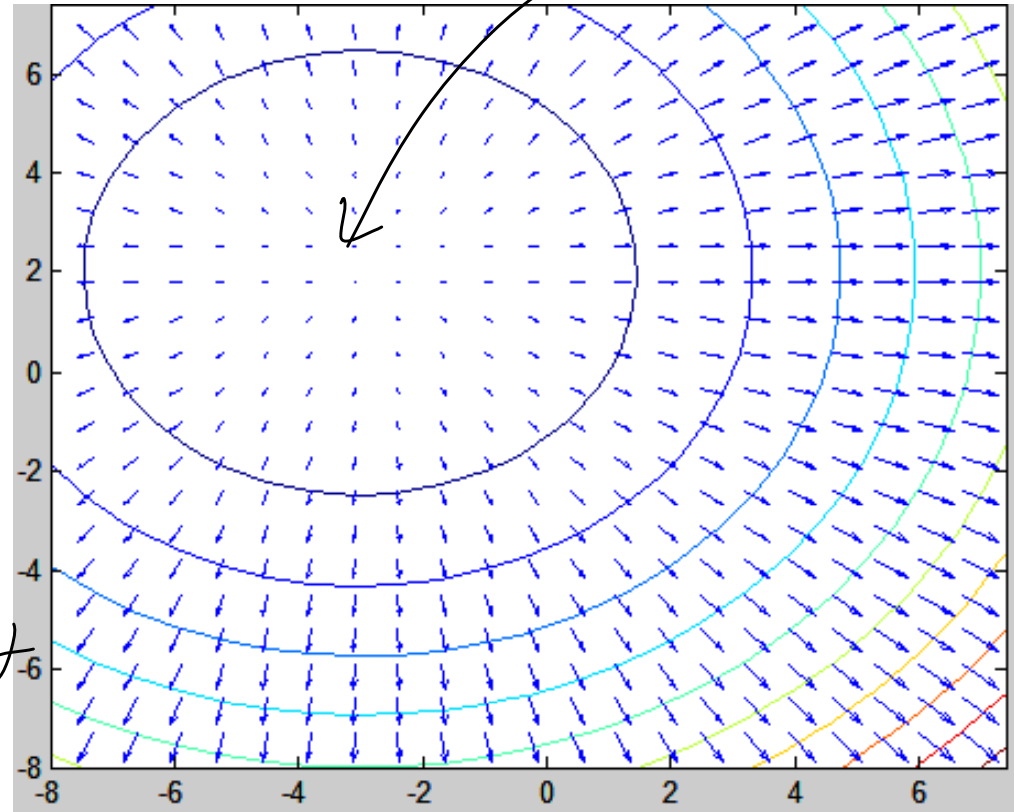
# Potential functions – basic idea

energy function over  $C_{free}$  (obstacle free configuration space)

Ideally  
global  
minimum  
at  
goal



gradient



# Definitions

$q \in \mathbb{R}^n$   
configuration

- Potential function

$$U : \mathbb{R}^n \rightarrow \mathbb{R}$$

- Gradient

$$\nabla U(q) = \begin{bmatrix} \frac{\partial U}{\partial q_1}(q) \\ \vdots \\ \frac{\partial U}{\partial q_n}(q) \end{bmatrix}$$

- Control

$$\dot{q} = -\nabla U(q)$$

# Attractive force = go to goal

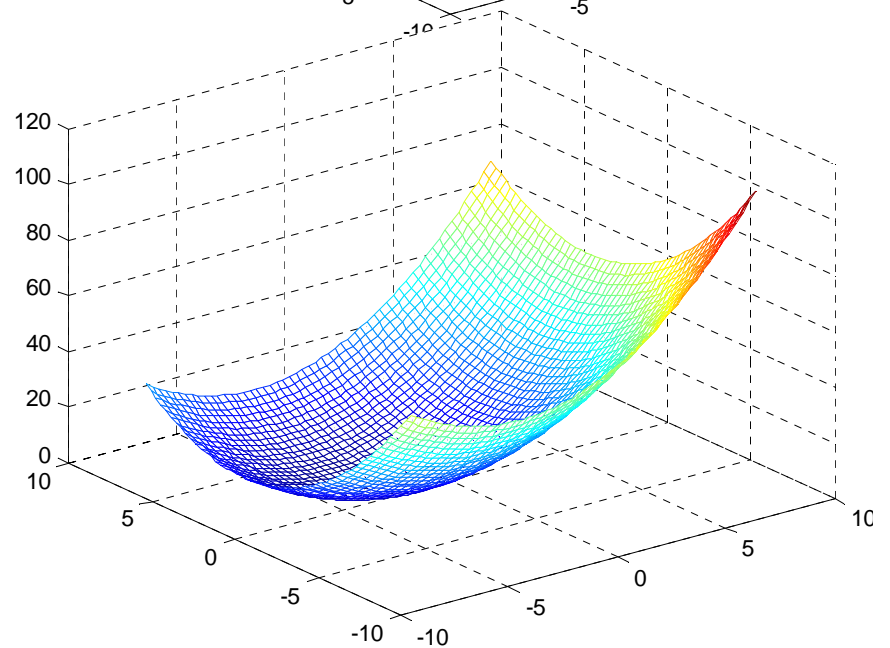
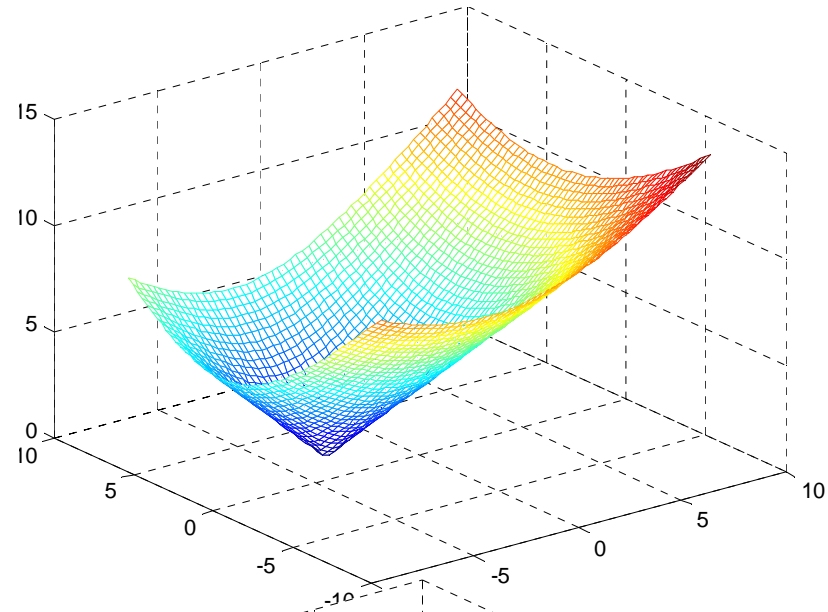
$$d(q, q_{goal}) = \text{norm}(q - q_{goal})$$

$$U_{att} = C \cdot d(q, q_{goal})$$

$$\nabla U_{att} = \frac{C}{d(q, q_{goal})} \cdot (q - q_{goal})$$

$$U_{att} = \frac{1}{2} C \cdot d(q, q_{goal})^2$$

$$\nabla U_{att} = C \cdot (q - q_{goal})$$



# Repulsive force = keep away from obstacles

Distance from obstacle

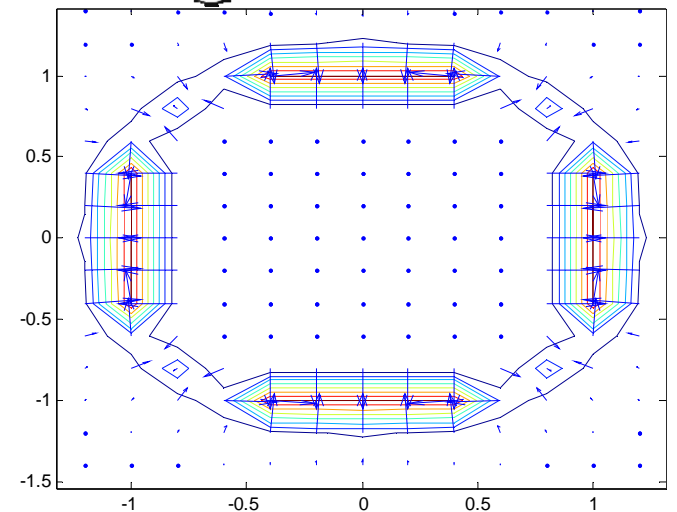
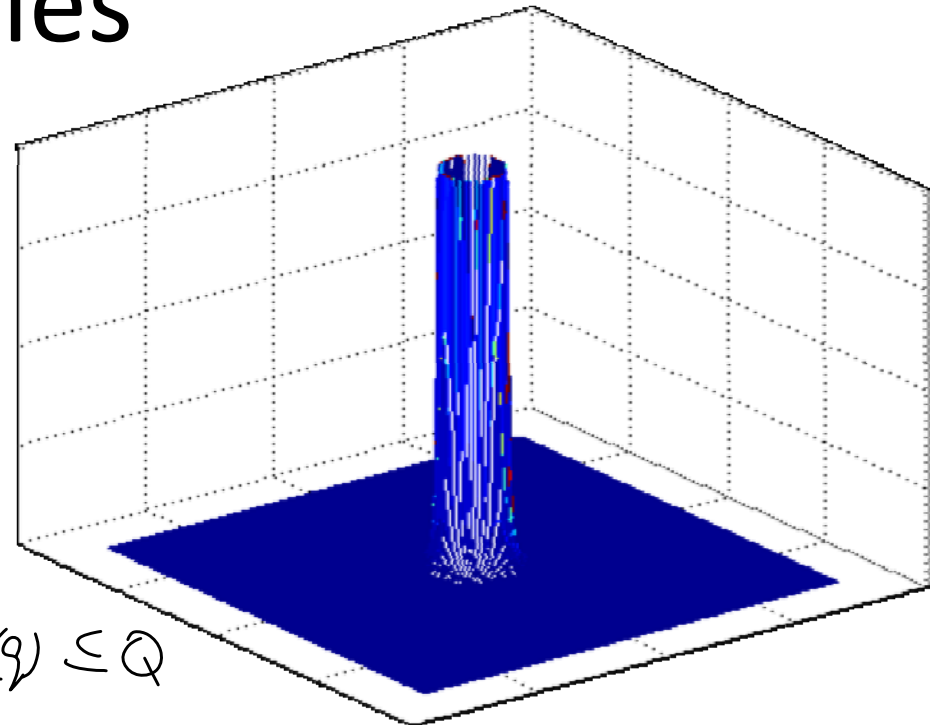
$$d_i(q) = \min_{q^* \in \text{obs};} d(q, q^*)$$

$$U_{\text{rep};} = \begin{cases} \frac{1}{2} \left( \frac{1}{d_i(q)} - \frac{1}{Q} \right)^2 \\ 0 \end{cases}$$

$$d_i(q) \leq Q$$

$$d_i(q) > Q$$

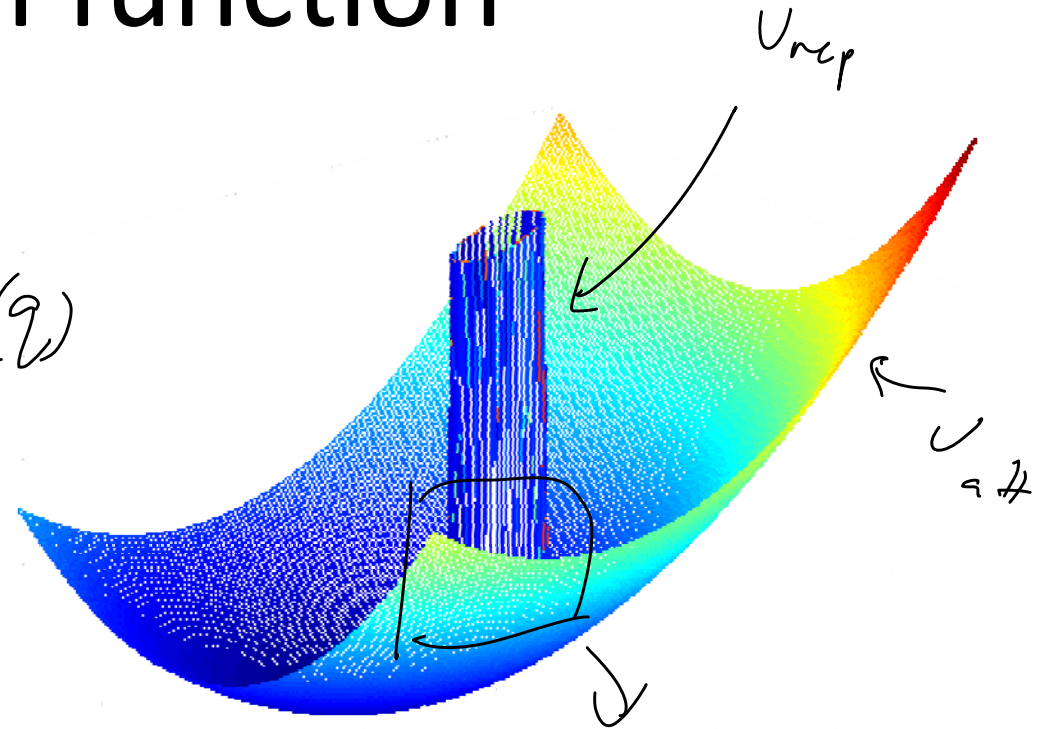
$$Q > 0$$





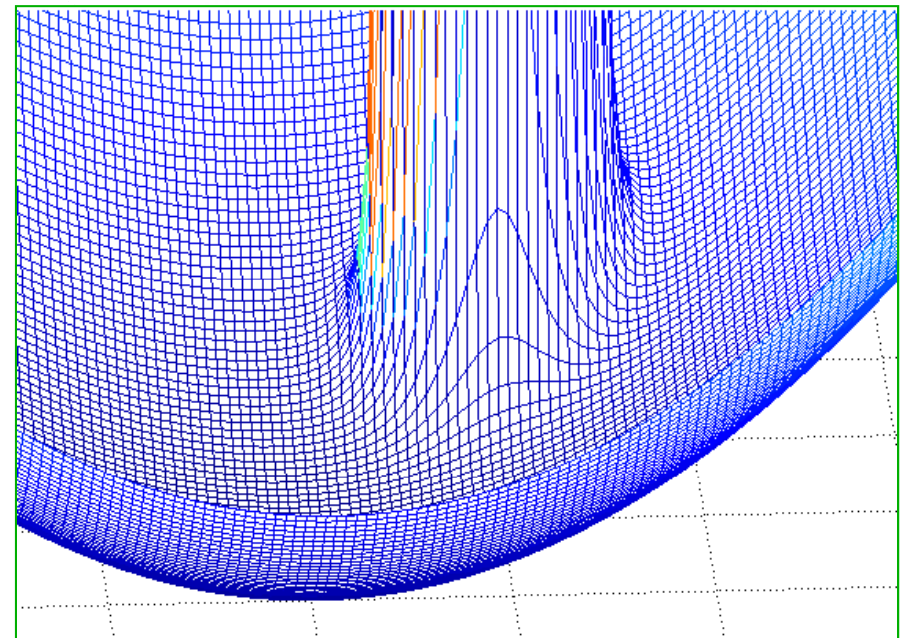
# Potential function

$$U(q) = U_{att} + \sum_{obs_i} U_{rep_i}(q)$$



pro

complete (every initial point will reach the goal\*)



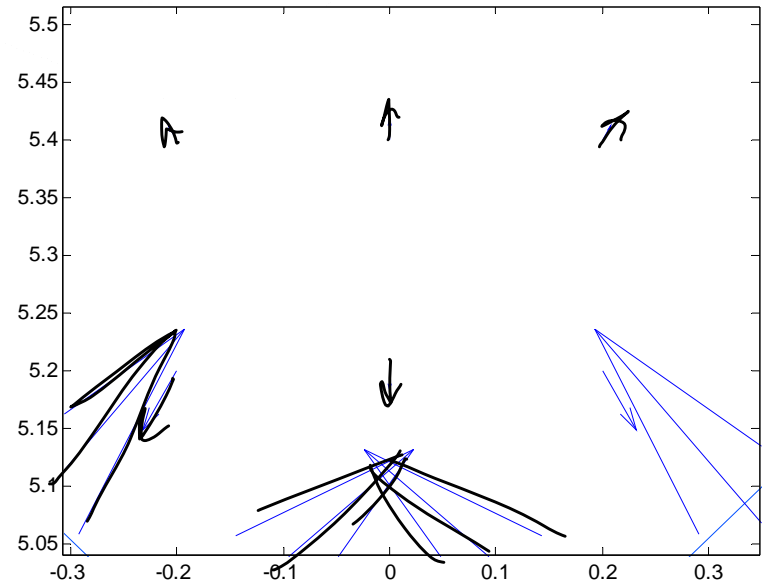
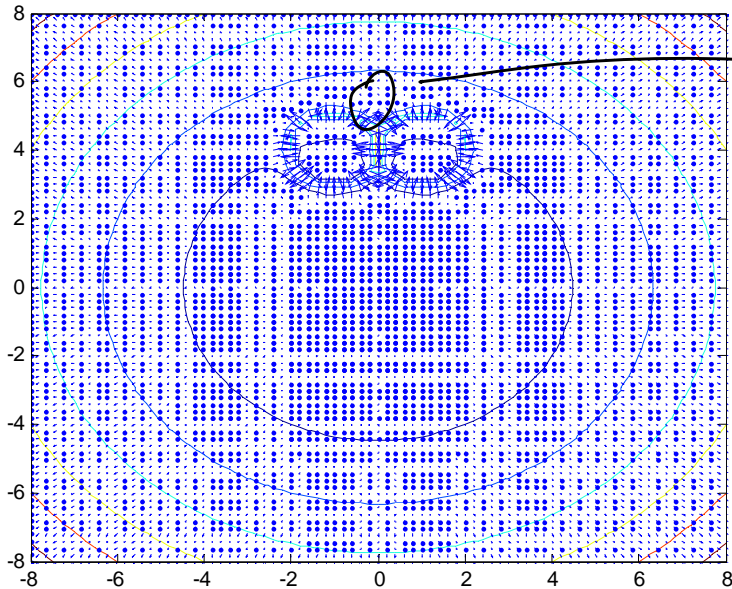
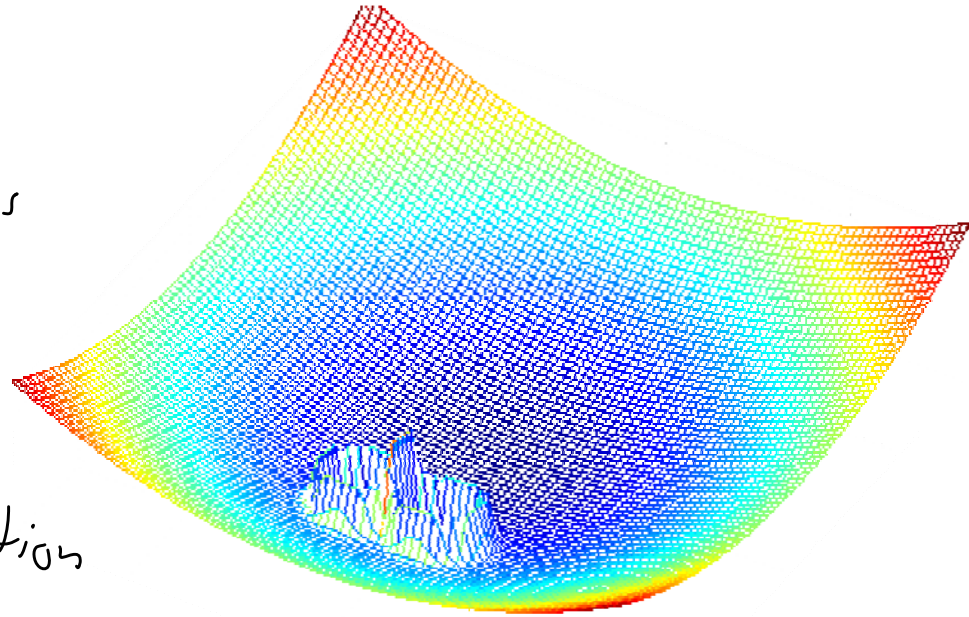
# Problem

local minima!

solutions: - Navigation functions

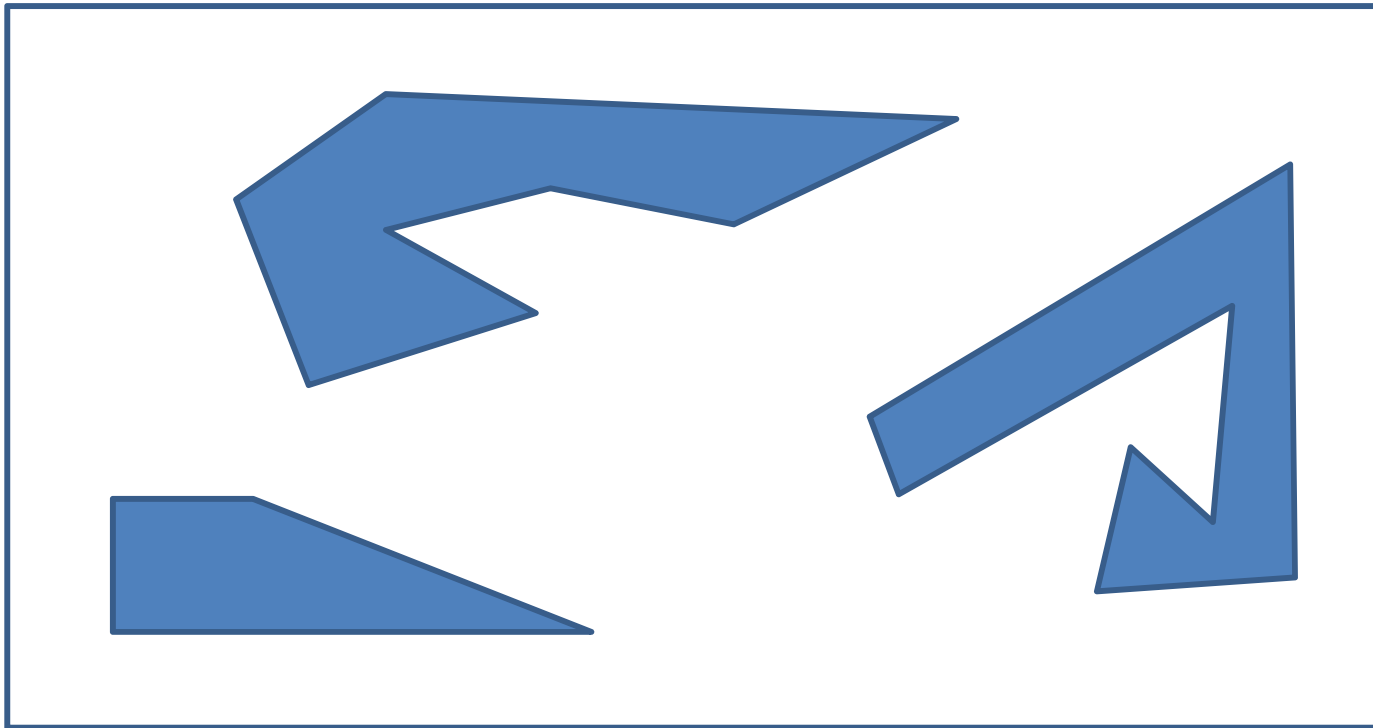
- potential functions  
over cell

decomposition

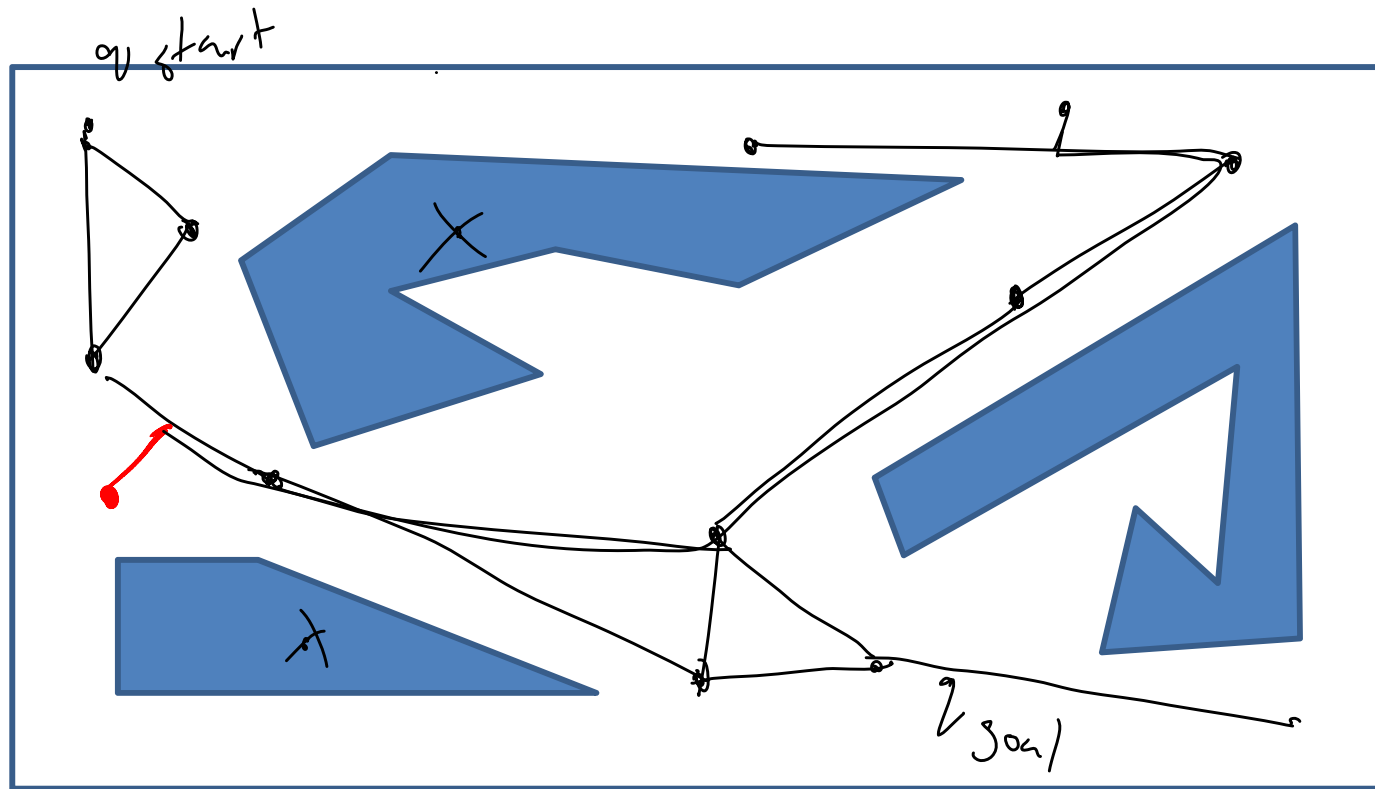


# Problem (2)

Complex environment



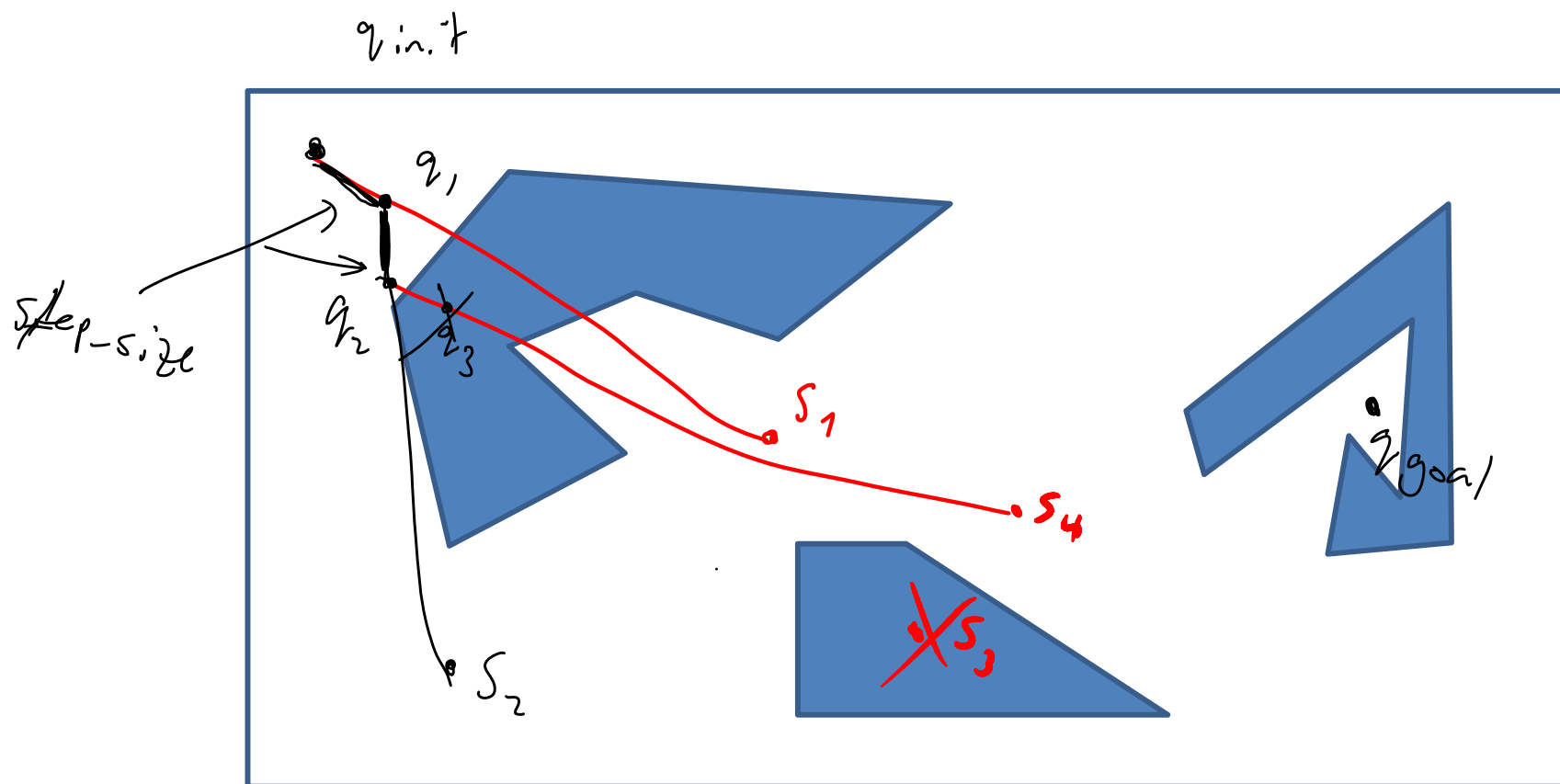
# Different approach - samples



- Probabilistically\resolution complete
- Good for complex configuration spaces



# Rapidly-Exploring Random Trees (RRT)



# RRTs

## Algorithm:

Given:  $q_{start}$ ,  $q_{end}$ ,  $step\_size$ ,  $n = \#$  of attempts to grow the tree

Find:  $G = (V, E)$   $V \in \mathbb{R}^n$   $E \in \mathbb{R}^n \times \mathbb{R}^n$

Init:  $V = \{q_{start}\}$   $E = \emptyset$

For  $i=1:n$

- sample  $q_{rand} \in C_{free}$

- find  $q_{near} =$  closest point  $q \in V$  to  $q_{rand}$

- generate  $q_{new}$ : point on line  $(q_{rand}, q_{near})$

that is  $step\_size$  away from  $q_{near}$

- if  $q_{new} \in C_{free}$  AND  $(q_{near}, q_{new}) \in C_{free}$

then  $V = V \cup \{q_{new}\}$ ,  $E = E \cup \{(q_{near}, q_{new})\}$



how to sample?

how to choose?

- try to connect  $q_{new}$  to  $q_{end}$   
if successful  $\rightarrow$  done!