### Potential Field Path Planning

### Reference:

## Principles of Robot Motion H. Choset et.al. Mit Press

Also: Siegwart text, section 6.3.2

### Potential Field Path Planning

- Simple idea: Have robot "attracted" to the goal and "repelled" from the obstacles
- Think of robot as a positively charged particle moving towards negatively charged goal – attractive force
- Obstacles have same charge as robot repelling force
- States far away from goal have large potential energy, goal state has zero potential energy
- Path of robot is from state of high energy to low (zero) energy at the goal
- Think of the planning space as an elevated surface, and the robot is a marble rolling "downhill" towards the goal



# Potential Field Path Planning

- A potential function is a function that may be viewed as energy
- the gradient of the energy is force
- Potential function guides the robot as if it were a particle moving in a gradient field.
- Analogy: robot is positively charged particle, moving towards negative charge goal
- Obstacles have "repulsive" positive charge

- Potential functions can be viewed as a landscape
- Robot moves from high-value to low-value Using a "downhill" path (i.e negative of the gradient).
- This is known as gradient descent –follow a functional surface until you reach its minimum

### **Potential Field**



- Attractive Potential Function is distance from goal
- High energy away from goal, Zero at goal
- Path is negative gradient, largest change in energy



**Figure 4.2** Different types of critical points: (Top) Graphs of functions. (Bottom) Gradients of functions.



**Figure 4.3** (a) A configuration space with three circular obstacles bounded by a circle. (b) Potential function energy surface. (c) Contour plot for energy surface. (d) Gradient vectors for potential function.

## **Computing Attractive Potential**

- With point robot, we can use distance as a measure, and velocity becomes the gradient
- Simplest: scaled distance to goal. Gradient becomes a constant, undefined at goal
- Better: Use a continuously differentiable function e.g. Quadratic Function of distance:

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

with the gradient

$$\begin{split} \nabla U_{\text{att}}(q) &= \nabla \left( \frac{1}{2} \zeta d^2(q, q_{\text{goal}}) \right), \\ &= \frac{1}{2} \zeta \nabla d^2(q, q_{\text{goal}}), \\ &= \zeta(q - q_{\text{goal}}), \end{split}$$



**Figure 4.4** (a) Attractive gradient vector field. (b) Attractive potential isocontours. (c) Graph of the attractive potential.

# **Computing Repulsive Potential**

- Strength of repulsive force should increase as we near the obstacle
- Compute potential in terms of distance to closest obstacle
- Multiple obstacles: compute repulsive potential over all obstacles

$$U_{\rm rep}(q) = \begin{cases} \frac{1}{2}\eta \left(\frac{1}{D(q)} - \frac{1}{Q^*}\right)^2, & D(q) \le Q^*, \\ 0, & D(q) > Q^*, \end{cases}$$

whose gradient is

$$\nabla U_{\text{rep}}(q) = \begin{cases} \eta \left( \frac{1}{Q^*} - \frac{1}{D(q)} \right) \frac{1}{D^2(q)} \nabla D(q), & D(q) \le Q^*, \\ 0, & D(q) > Q^*, \end{cases}$$



Figure 4.5 The repulsive gradient operates only in a domain near the obstacle.

#### **Total Potential Function**



16-735, Howie Choset, with slides from Ji Yeong Lee, G.D. Hager and Z. Dodds

### **Potential Field**



- Obstacles create high energy barriers
- Gradient descent follows energy minimization path to goal

#### **Gradient Descent**

Gradient Descent:

- $q(0)=q_{start}$
- -i = 0
- − while  $|| \nabla U(q(i)) || > ε$  do
  - $q(i+1) = q(i) \alpha(i) \nabla U(q(i))$
  - i=i+1





Local minimum: attractive force (goal) = repulsive force (obstacles)

### **Potential Field Methods**



Local minimum: attractive force = repulsive force Solution: Take a random walk – perturb out of minima Need to remember where you have been!

## **Online** Distance Computation



Figure 4.7 Local minima of rays determine the distance to nearby obstacles.

### **Potential Fields Summary**

- More than just a path planner: Provides simple control function to move robot: gradient descent
- Allows robot to move from wherever it finds itself
- Can get trapped in local minima
- Can be used as online, local method:
  - As robot encounters new obstacles compute the Potential Function onlir
  - Laser/sonar scans give online distance to obstacles

