

# Potential Field Methods

- Idea robot is a particle
- Environment is represented as a potential field (locally)
- Advantage - capability to generate on-line collision avoidance

Compute force acting on a robot - incremental path planning

$$F(q) = -\nabla U(q)$$

Example: Robot can translate freely , we can control independently  
Environment represented by a potential function

$$U(x, y)$$

Force is proportional to the gradient of the potential function

$$\begin{bmatrix} \dot{x} \\ \dot{y} \end{bmatrix} = -\nabla U(x, y)$$

Some slide thanks to <http://cs.cmu.edu/~motionplanning>

## Attractive potential field

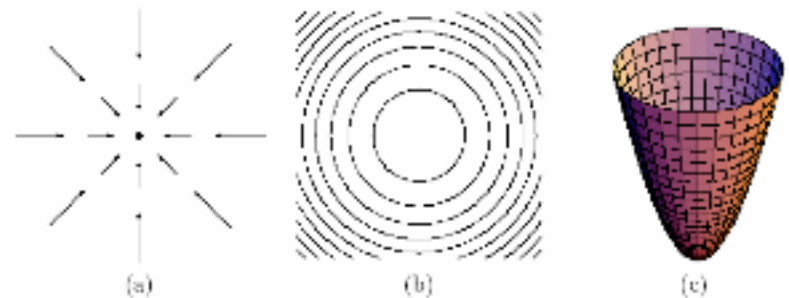
- Linear function of distance

$$U_a(q) = \xi \|q - q_{goal}\| \quad F_a(q) = -\nabla U_a(q) = -\xi \frac{(q - q_{goal})}{\|q - q_{goal}\|}$$

- Quadratic function of distance

$$U_a(q) = \xi \frac{1}{2} \|q - q_{goal}\|^2 \quad F_a(q) = -\nabla U_a(q) = -\xi (q - q_{goal})$$

Combination of two - far away use linear  
closer by use parabolic well



## Repulsive potential field

$$U_{rep} = \frac{1}{2} \nu \left( \frac{1}{\rho(q, q_{obst})} - \frac{1}{\rho_0} \right)^2 \quad \text{if} \quad \rho(q, q_{obst}) \leq \rho_0$$

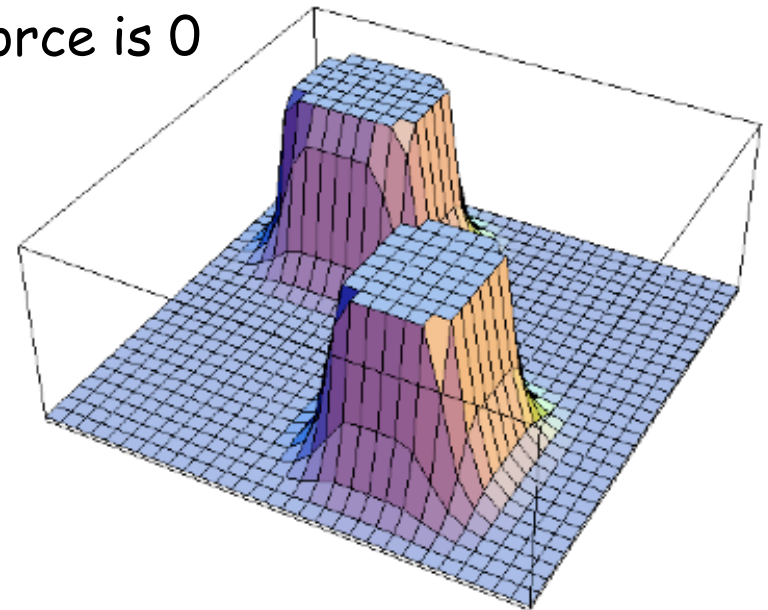
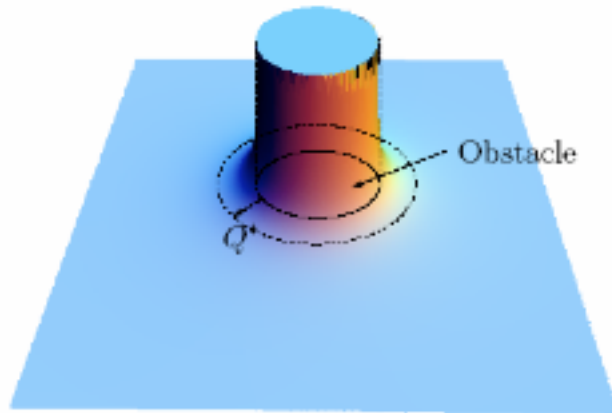
$\nearrow$  Minimal distance between the robot and the obstacle

$$\text{else} \quad U_r(q) = 0$$

Minimal distance between the robot and the obstacle

$$F_{rep} = -\nabla U_{rep} = \nu \left( \frac{1}{\rho(q)} - \frac{1}{\rho_0} \right) \frac{1}{\rho(q)^2} \frac{q - q_{obs}}{\rho(q)}$$

Outside of sensitivity zone repulsive force is 0



## Another example of repulsive potential field

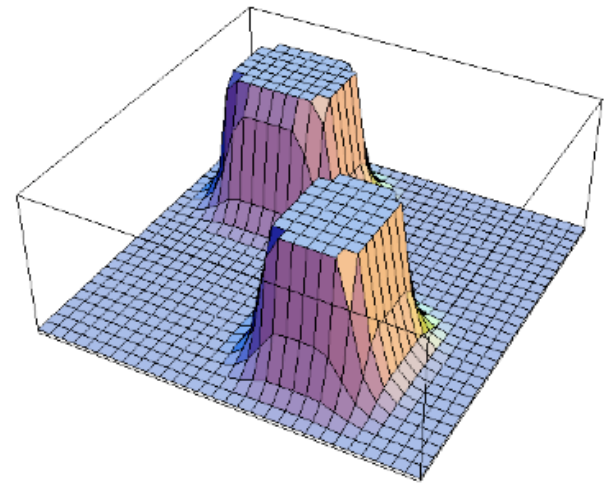
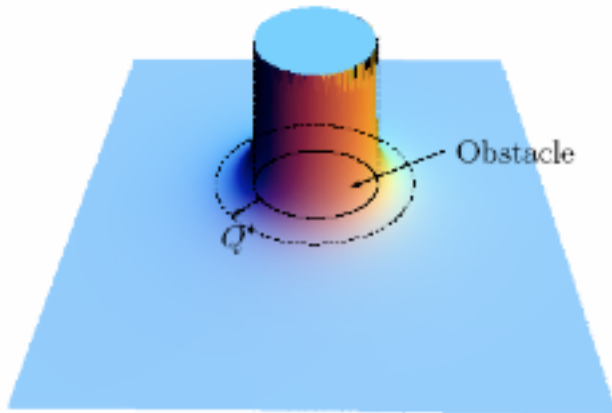
$$U_r(q) = \frac{1}{2}\nu \left( \frac{1}{\rho(q, q_{obst})} - \frac{1}{\rho_0} \right) \quad \text{if} \quad \rho(q, q_{obst}) \leq \rho_0$$

$\nearrow$

$$\text{else} \quad U_r(q) = 0$$

Minimal distance between the robot and the obstacle

Previously - repulsive potential related to the square of the Inverse distance - here just proportional to inverse distance  
Note: need to compute gradient to get the force

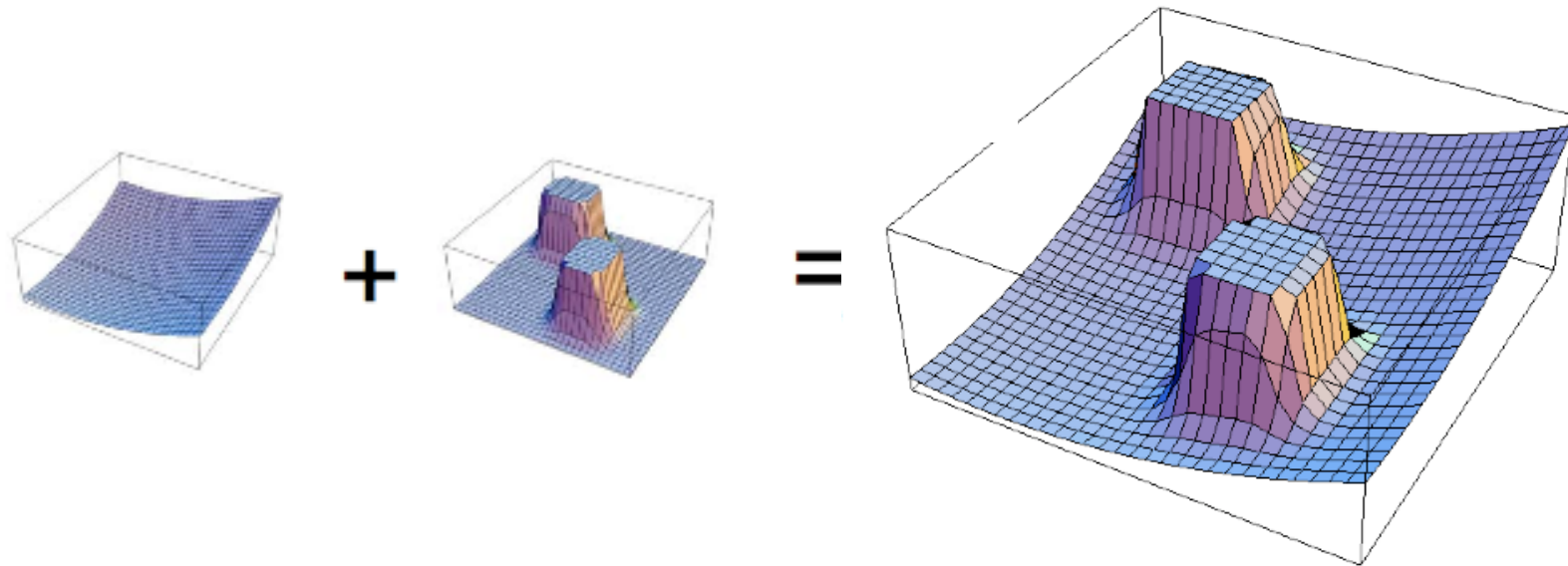


# Potential Function

Resulting force

$$F(q) = -\nabla(U_a(q) + U_r(q))$$

Iterative gradient descent planning  $q_{i+1} = q_i + \delta_i \frac{F(q)}{\|F(q)\|}$

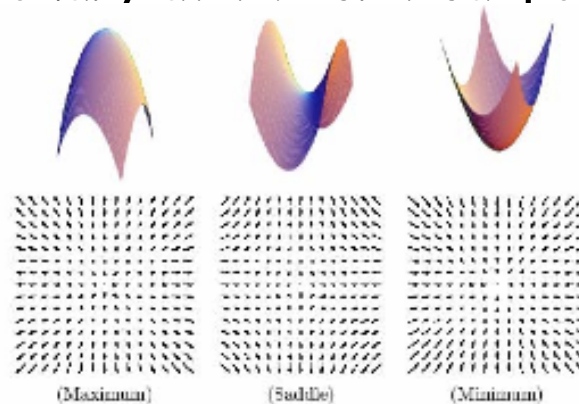


# Potential Fields

- Simple way to get to the bottom, follow the gradient

$$\begin{bmatrix} \dot{x} \\ \dot{y} \end{bmatrix} = -\nabla U(x, y) \qquad \dot{q} = -\nabla U(q)$$

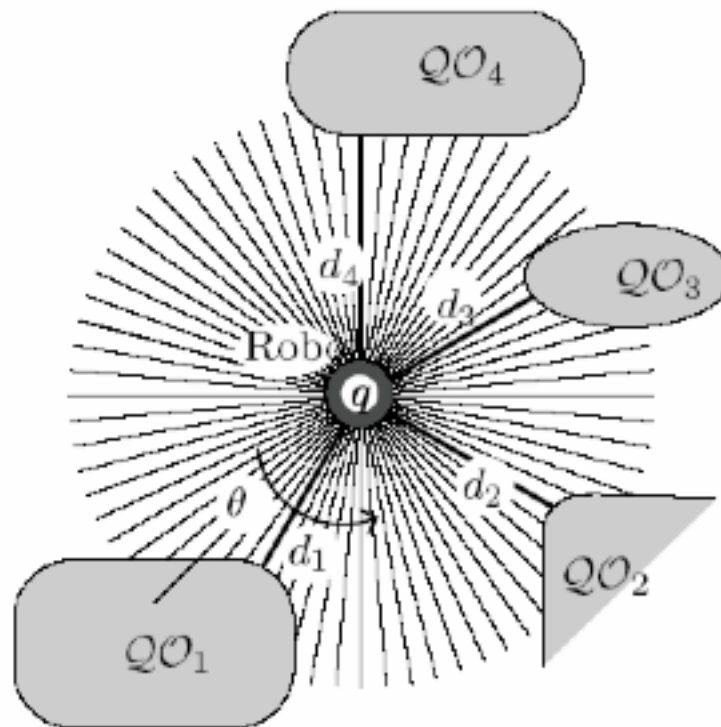
- i.e. Gradient descent strategy  $q_{i+1} = q_i + \delta_i \frac{F(q)}{\|F(q)\|}$
- A critical, stationary point is such that  $\nabla U(q) = 0$
- Equation is stationary at the critical point



- To check whether critical point is a minimum - look at the second order derivatives (Hessian for m -> n function)

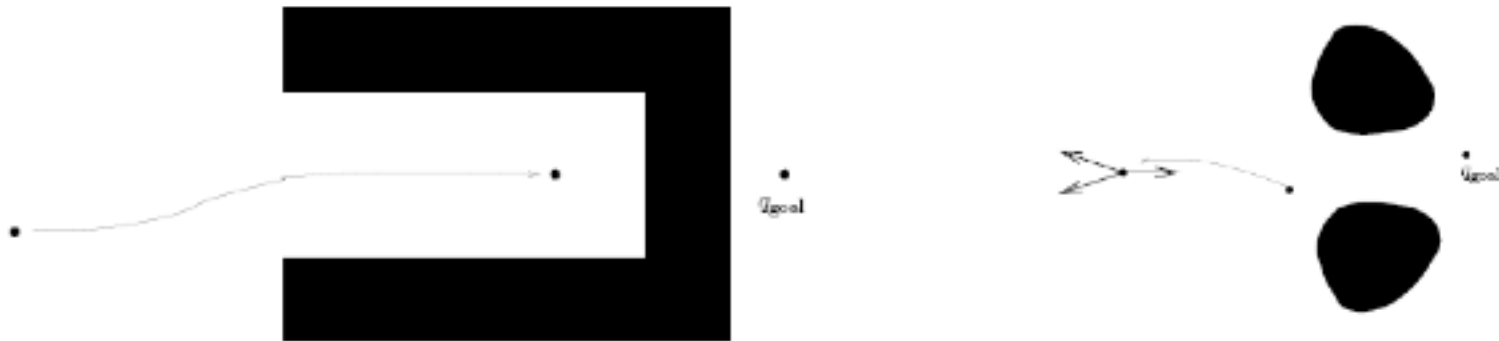
# Computing Distances

- Using sensor measurements



# Potential Functions

- How do we know we have single global minimum ?



- If global minimum is not guaranteed, need to do something else then gradient descent
- Design functions in such a way that global minimum can be guaranteed



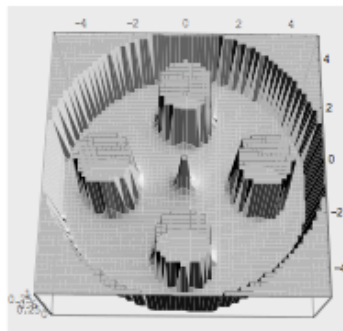
# Potential function

- Heuristics for escaping the local minima
- Can be used in local and global context
- Numerical techniques, Random walk methods
- Navigation functions (Rimon & Kodistchek, 92)
- Navigations in sphere worlds and worlds diffeomorphic to them

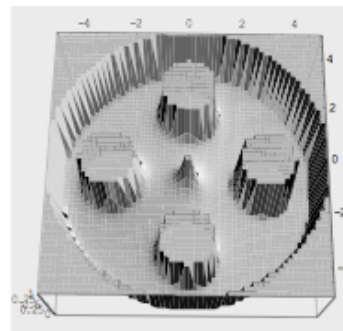
# Navigation functions

$$\phi(q) = - \frac{d^2(q, q_{goal})}{[d(q, q_{goal})^{2k} + \beta(q)]^{1/k}} \quad \beta(q) \quad \text{Obstacle term}$$

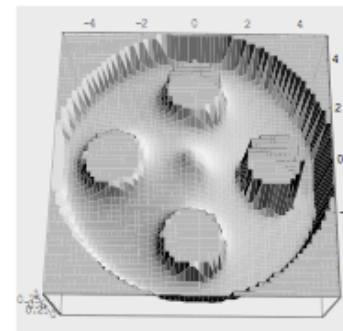
- For sufficiently large  $k$  - this is a navigation function [Rimon-Koditschek, 92]



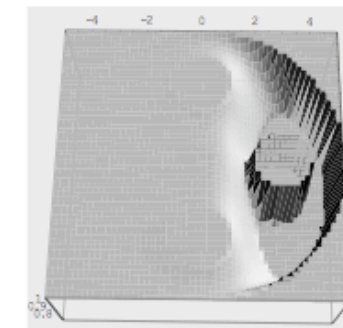
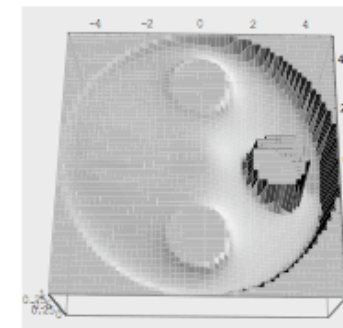
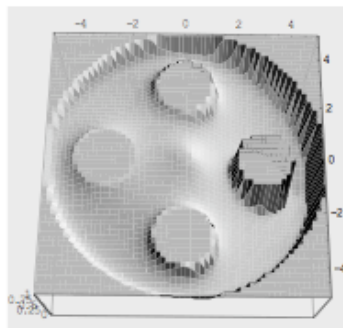
$k=3$



$k=4$

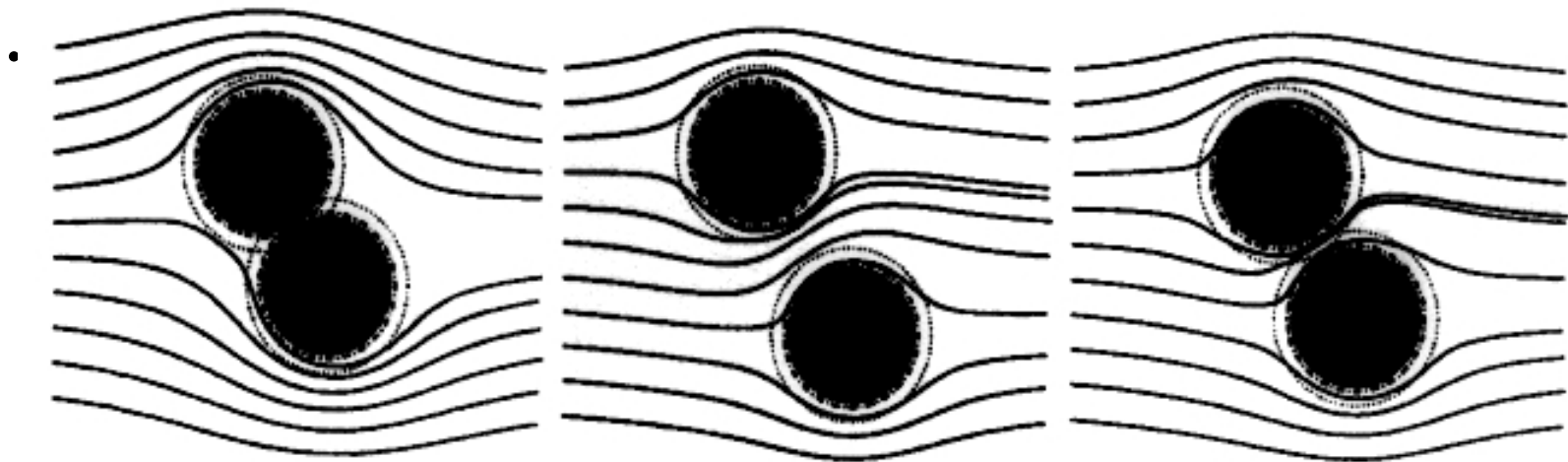


$k=6$



## Potential Field Path Planning: Using Harmonic Potentials

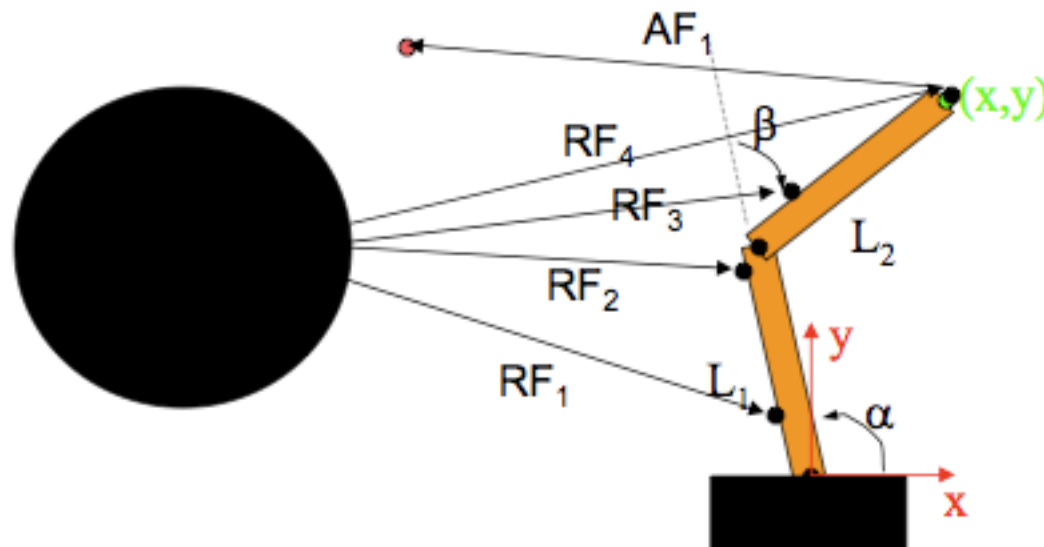
- Hydrodynamics analogy
  - robot is moving similar to a fluid particle following its



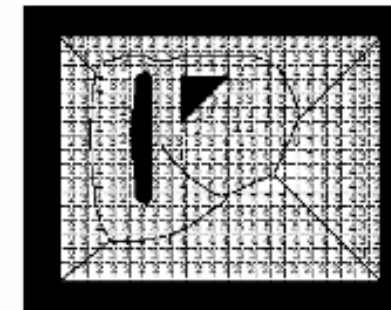
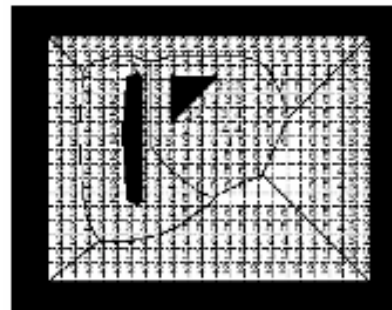
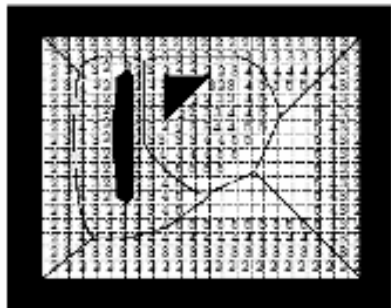
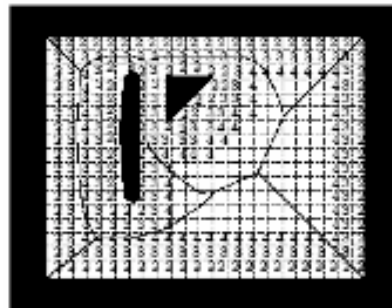
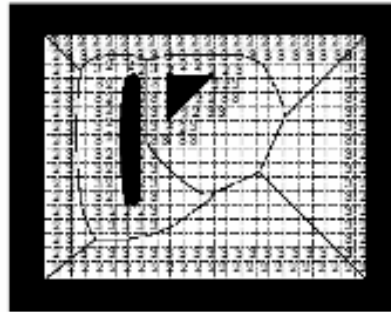
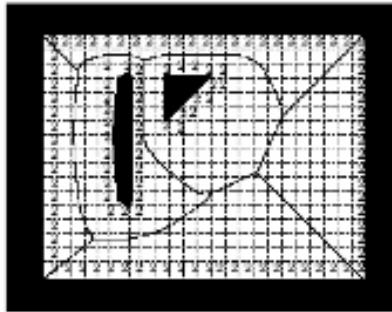
- Note:
  - Complicated, only simulation shown

# Potential fields for Rigid Bodies

- So far robot was considered a point- gradient of the potential function - force acting on a point
- how to generalize to manipulators of objects ?
- Idea - forces acting on objects - forces acting on multiple points of the object (black board)
- <http://www.cs.cmu.edu/~motionplanning/>
- For robots, pick enough control points to pin down the robot - define forces in workspace - map them to configuration space



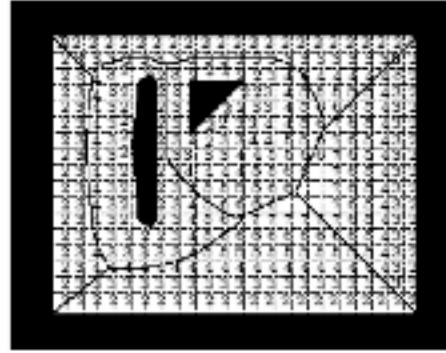
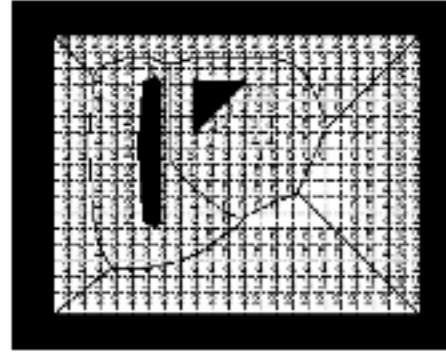
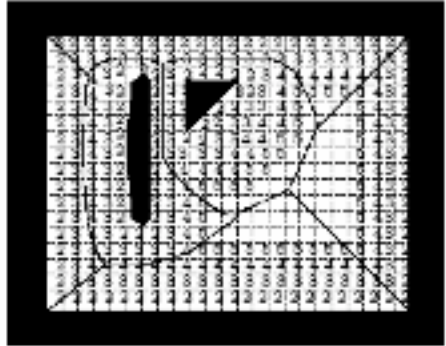
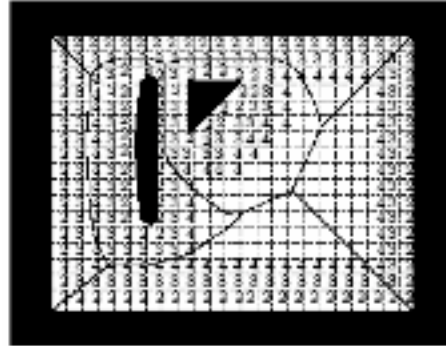
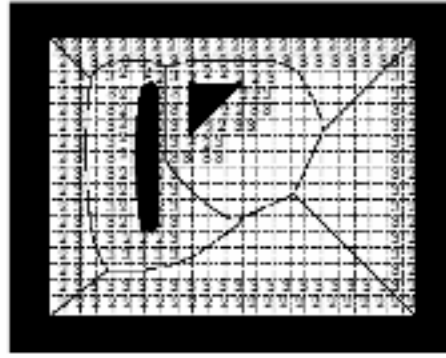
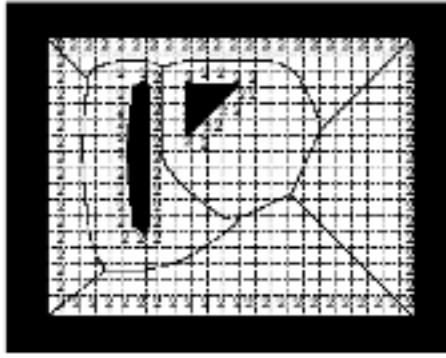
# Brush fire algorithm



Create a queue  $Q$  of all pixels at the boundary of obstacles  
For each, set the boundary to 1  
And the free space to 0.

- For each element in the  $Q$
- If  $d(q) = 0$  set  $d(q) = 1 + \min d(q')$  of the neighbours which differ from 0
- Add all neighbours to the  $Q$  with  $d(q) = 0$

Resulting map - distance to the nearest obstacle -  
gradient of the distance



# Wave-front planner

- Apply brushfire algorithm starting from the goal
- Label goal 2, add all zeros labels to Q
- While Q non-empty
- Set  $d(q) = 1 + \min d(q)$  of the neighbours which are different from 1 and 0
- Add all neighbours of the selected point to the Q
- The results is a distance for each point to the goal
- Follow gradient descent to the goal

# Navigation functions - Discrete version

- Useful potential functions - for any starting point you will reach a goal: Navigation function (see previously how to define these in continuous spaces)
- In discrete state space - navigation function has to have some value for each state, consider grid
- Goal: will have zero potential
- Obstacles have infinite potential
- Example of useful navigation function: optimal cost to go to goal  $G^*$  (example) , assuming that for each state

$$l(x, u) = 1$$



# Wavefront planner

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
4	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
3	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	2	
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
4	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
3	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	
2	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	0	0	0	0	0	0	3	3	
0	0	0	0	0	0	0	0	0	0	0	0	0	0	3	2	
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	1	1	1	1	1	1	1	1	0	0	0	0
3	0	0	0	0	1	1	1	1	1	1	1	1	5	5	5	5
2	0	0	0	0	0	0	0	0	0	0	0	0	5	4	4	4
1	0	0	0	0	0	0	0	0	0	0	0	0	5	4	3	3
0	0	0	0	0	0	0	0	0	0	0	0	0	5	4	3	2
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

7	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	0	7	7	7	7	7
4	0	0	0	0	1	1	1	1	1	1	1	1	6	6	6	6
3	0	0	0	0	1	1	1	1	1	1	1	1	5	5	5	5
2	0	0	0	0	0	0	0	0	0	0	7	6	5	4	4	4
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7	18	17	16	15	14	13	12	11	10	9	9	9	9	9	9
6	17	17	16	15	14	13	12	11	10	9	8	8	8	8	8
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1	17	16	15	14	13	12	11	10	9	8	7	6	5	4	3
0	17	16	15	14	13	12	11	10	9	8	7	6	5	4	2
	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14

- Zeros should exist only in unreachable regions exist
- Ones for regions which cannot be reached
- Other book keeping methods can be used

# Finding the path

- From any initial grid cell, move toward the cell
- with the lowest number - follow the steepest
- descent

