

# LECTURE 13

**Schema theory**  
**Reactive paradigm**

**Ola Ringdahl**



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# CONTENTS OF THIS LECTURE

- Biological Foundations of Robot Behavior (part 2)
  - Schema theory
- Reactive paradigm
  - Subsumption
  - Potential fields



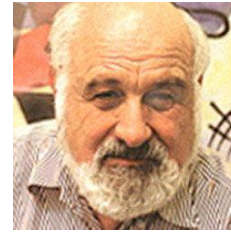
# SCHEMA THEORY



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# SCHEMA THEORY

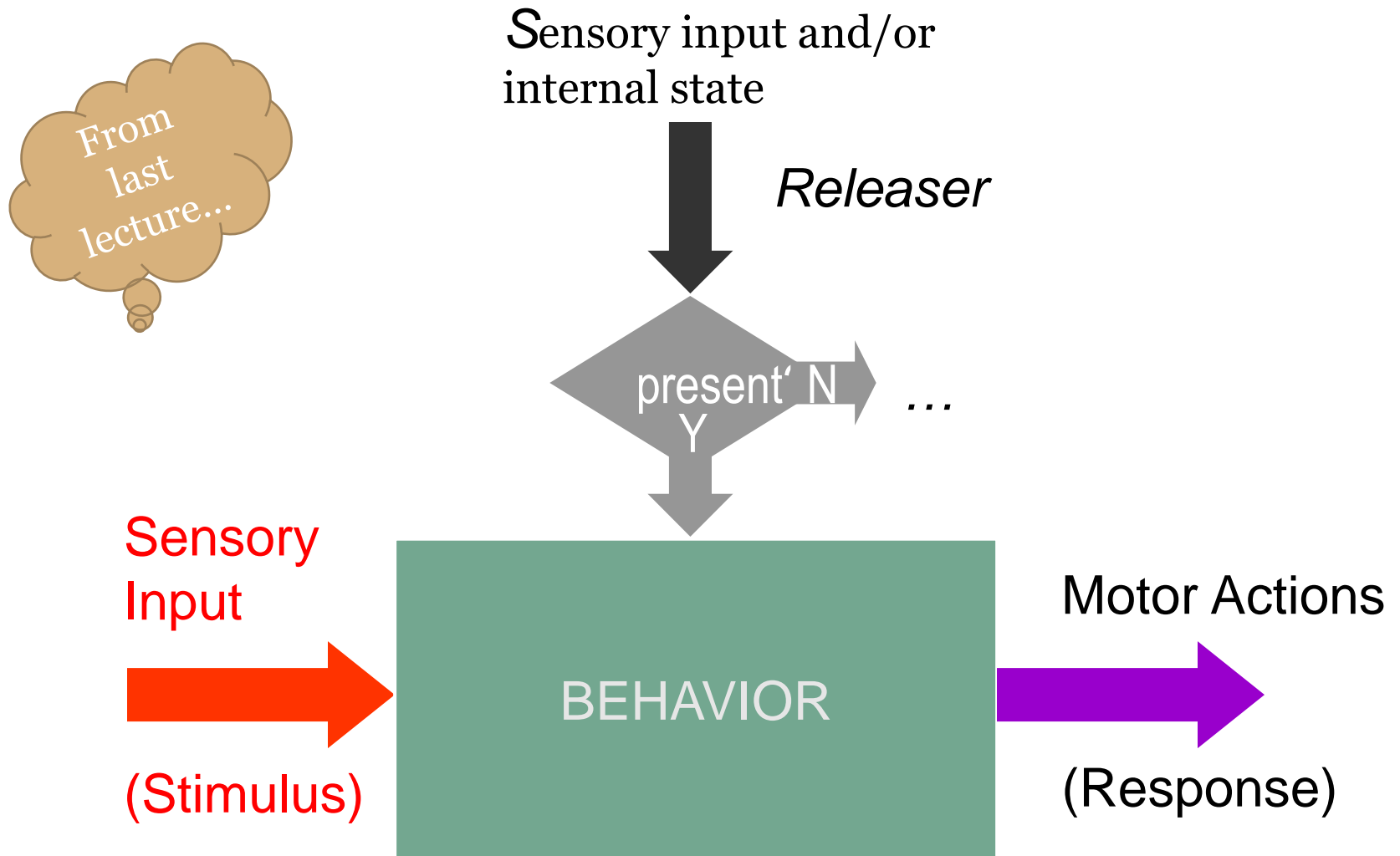
- A way to describe and model behaviours
- Inspired by psychology (used there since the early 1900's)
- Applied to robotics by Michael Arbib



- A schema is a generic template for performing an activity. It contains (compare with OOP):
  - Models and algorithms for how to act and/or perceive
  - Data structures (not in purely reactive behaviours)

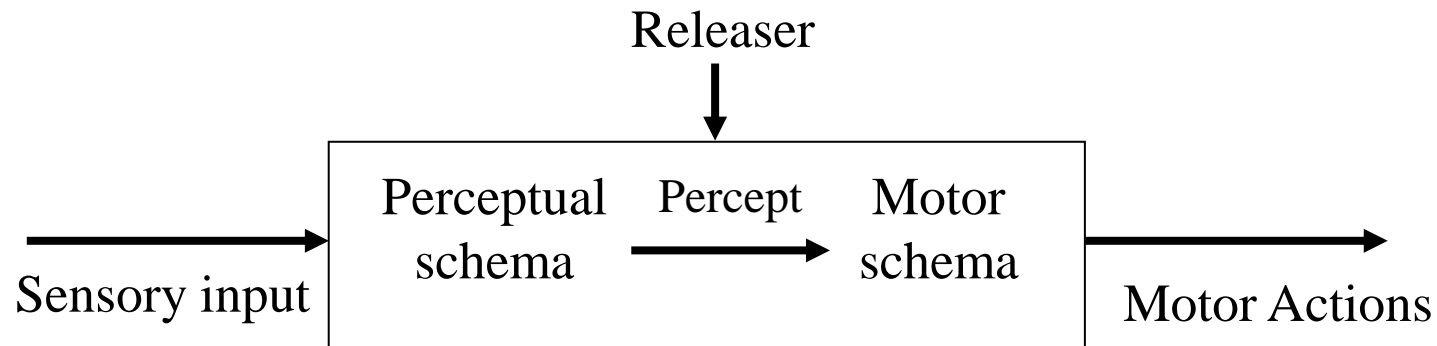


# INNATE RELEASING MECHANISMS (IRM)



# BEHAVIORS AS SCHEMAS

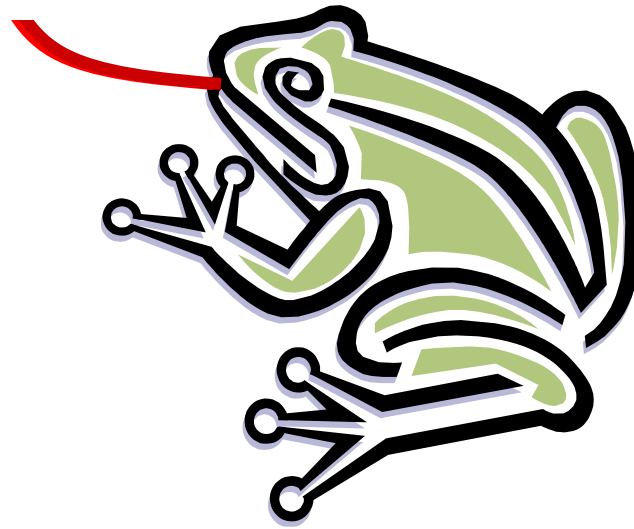
- The Perceptual Schema produces a Percept
- The Percept is read by the Motor Schema
- Motor actions are vectors: direction and strength



- Several schemas may be active at the same time
- Actions are combined by vector addition

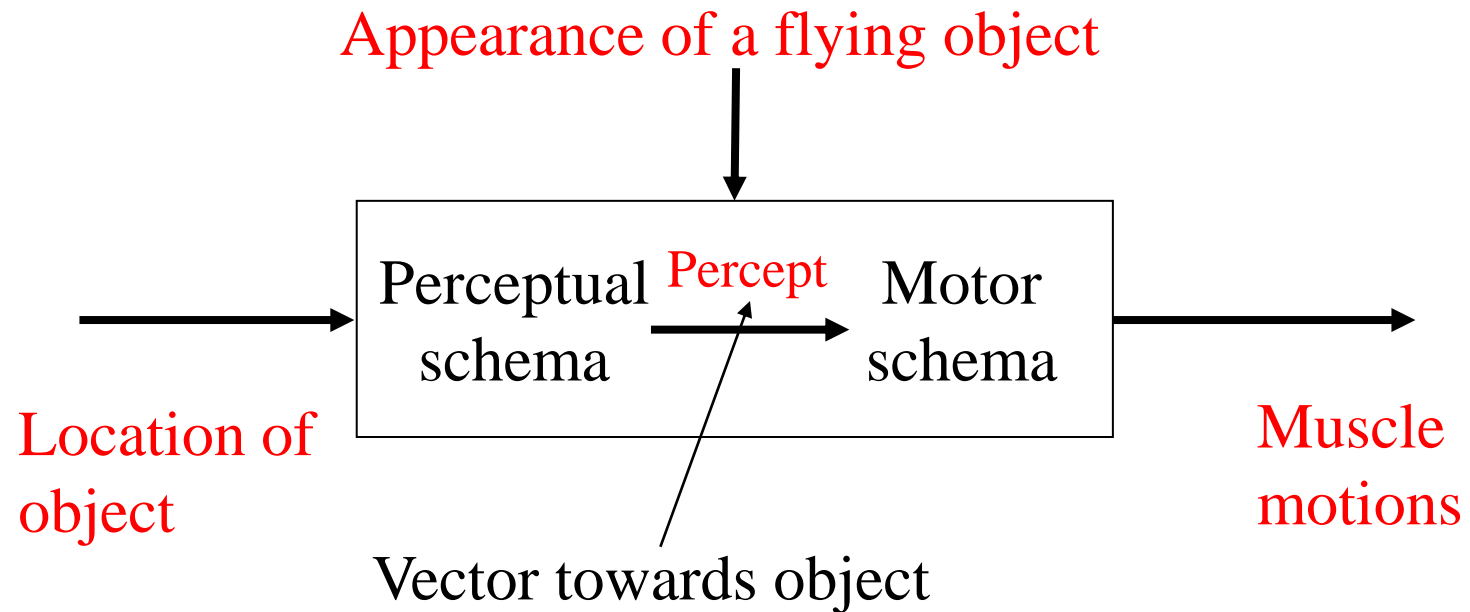


# Fly Snapping Behavior



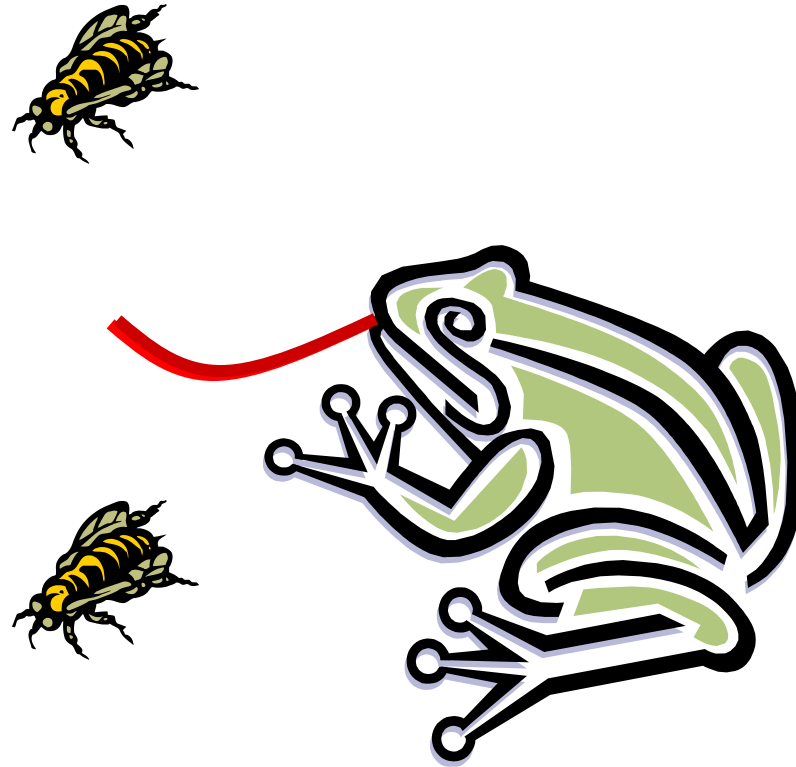
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# Fly Snapping Behavior





2 flies at the same time



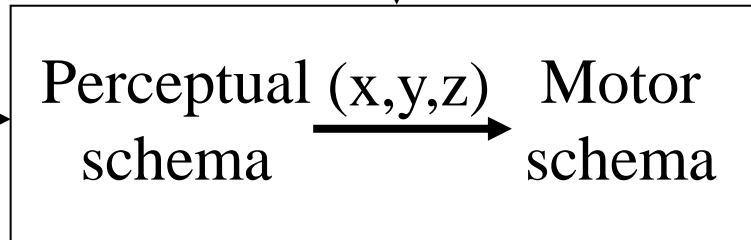
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# 2 flies at the same time

Appearance of flying object 1



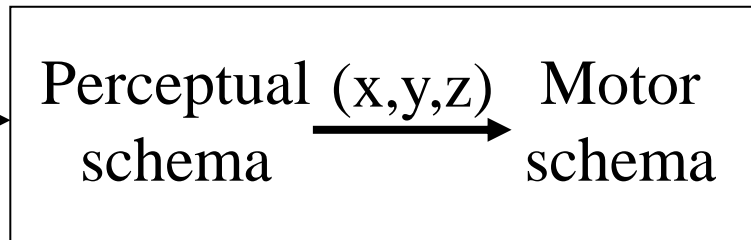
Location of  
object 1



Appearance of flying object 2



Location of  
object 2



Vector  
summation



Muscle  
motion

The toad will indeed snap in the middle!

1 fly may be more than enough



# CONCLUSIONS FOR ROBOTICS

- Maybe programming in terms of parallel behaviors is better than STRIPS or other deliberative approaches
- Individual behaviors
  - are often simple
  - are independent
  - can be implemented as threads
- Sequences of behaviors
  - can appear from parallel behaviors activated by releasers (internal states and/or external sensor data)
  - may appear intelligent for an observer
- The system “degrades gracefully” if any part fails since it’s modular
- IRMs and Schemas are nice ways to start thinking about the computational structure of programming a robot



# REACTIVE PARADIGM

**Mataric - Reactive paradigm (p163-169)**



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# THE REACTIVE PARADIGM

- Reacted against classical AI  
(the hierarchical paradigm/ deliberative systems)
- Reduced emphasis on:
  - Knowledge representation
  - Planning
- Increased emphasis on:
  - Sensing and acting within the environment
  - "Situatedness" - interaction with the environment is crucial for intelligent behavior
  - "Embodiment" – the structure of the body is tightly connected to behavior and intelligence.



# THE REACTIVE PARADIGM

**Suggested by Rodney Brooks MIT (1986)**

- “Planning is just a way of avoiding figuring out what to do next”
- “The world is its own model”
- “Elephants don’t play chess”
- Complex control systems are not necessary for complex behavior
- Intelligence is in the eye of the observer
- No representation. No calibration. No complex computers. No High bandwidth communication



# RELATED QUOTES

*"Everybody's got a plan until they get hit"*

Mike Tyson – unknowingly commenting on  
*The closed world assumption*



*"If you want to make God laugh, tell him  
about your plans"*

Woody Allen – general disbelief in  
predictions of the future

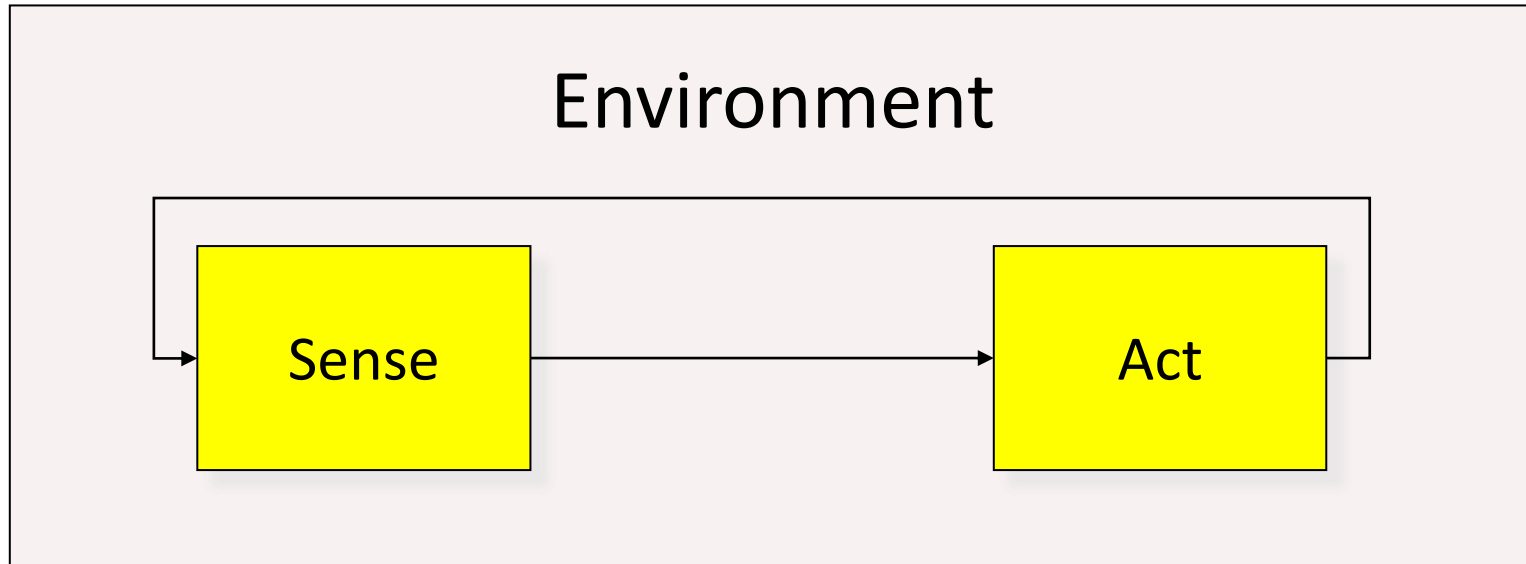


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# THE REACTIVE PARADIGM

Walter 1953, Breitenberg 1984, Brooks 1986



Method: “Sense-Act

- Sensor data is immediately transformed into Actions
- Based on biology (e.g. insects)
  - Stimuli → Response



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# THE REACTIVE PARADIGM

- The basic components are called *BEHAVIORS*
- Each behavior
  - is implemented as reactive rules, mapping stimuli to responses without thinking or computing
  - has no (or little) memory
  - computes responses continuously
- All behaviors exist in parallel and are activated (triggered) by percepts from the sensors
- *Coordination* if several behaviors are active at the same time



# ARCHITECTURAL ISSUES

- How to represent behaviors
- How to trigger behaviors
- How to coordinate behaviors

Two common architectures:

- Subsumption architecture
- Potential fields



# REACTIVE PARADIGM ALGORITHMS

**Subsumption**  
**Potential fields**



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# Subsumption Architecture

Suggested by Rodney Brooks MIT (1986)



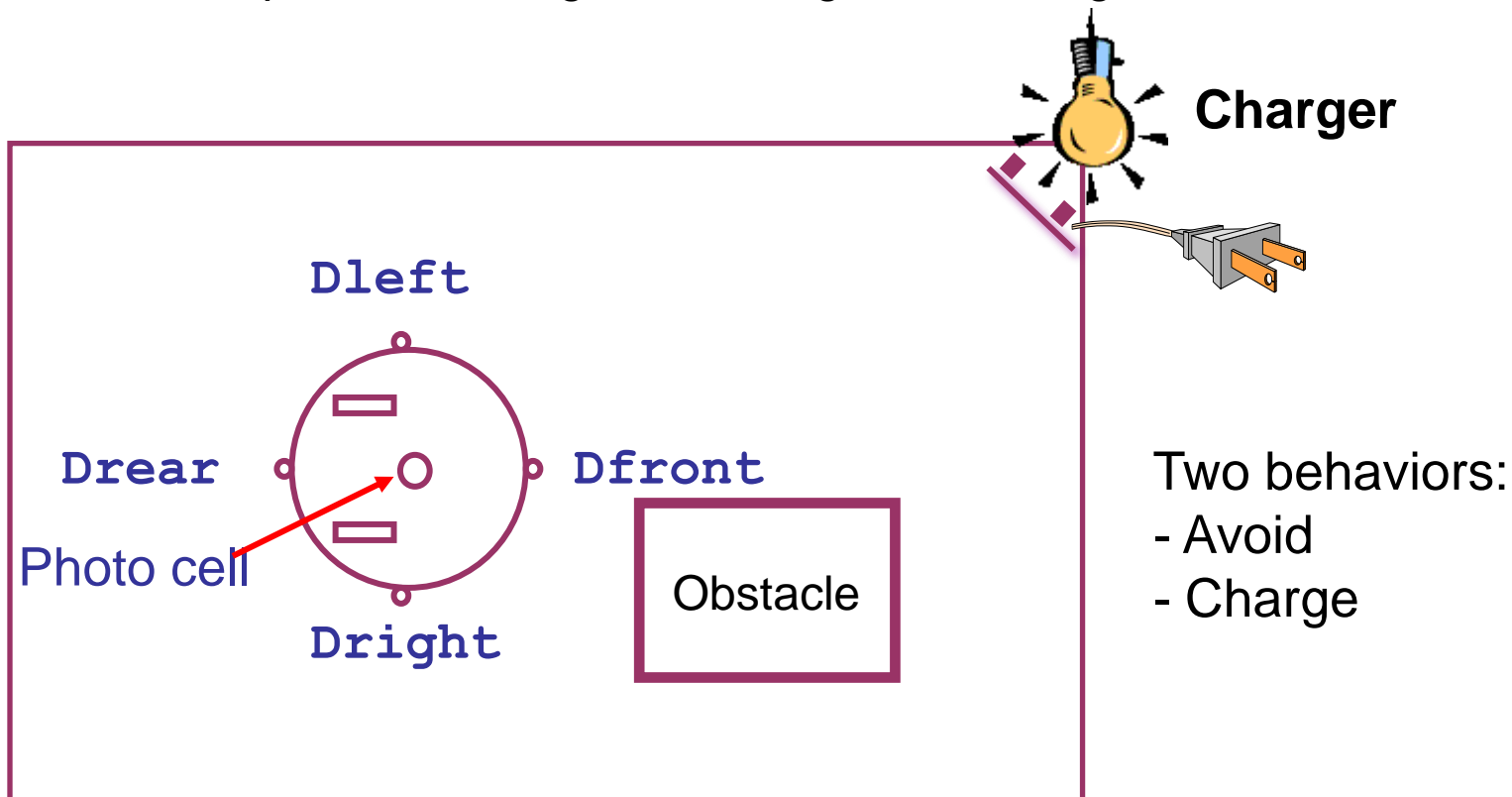
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# Example Subsumption Architecture



## A self-charging robot

- 2 engines for propulsion and steering
- Controlled by speed and angular velocity
- Battery voltage sensors  $Bat$
- 4 ultrasonic distance sensors  $D_{left}, D_{right}, D_{front}, D_{rear}$
- 1 photocell that gives the angle  $D$  to the light source



# Subsumption Architecture

## Avoid



The behavior *Avoid* contains code that maps the signals from the 4 US sensors to the actuator (Motors):

```

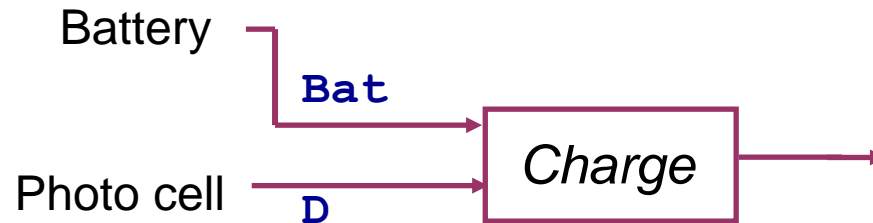
If Dfront<d0 or Drear<d0 or Dleft<d0 or Dright<d0
  % Act only if an obstacle is close
  if Drear<min(Dleft, Dright, Dfront) and Dfront>S0
    "move ahead"    % Go away from obstacles
  else
    if Dleft<Drigh then "turn right"
    if Drigh<Dleft then "turn left"
  end
end
end
  
```

Note: responses are computed continuously (don't wait until finished)



# Subsumption Architecture

## Charge



The behavior *Charge* contains code that turn the robot towards a light source (the charger), move forward and dock.

The PhotoCell senses light in direction *D* (*D* = 0 if light is straight forward)

For instance:

```
If Bat < b0 and  
    if D > Dtol then "turn right"  
    if D > -Dtol then "turn left"  
    "move ahead"
```

End

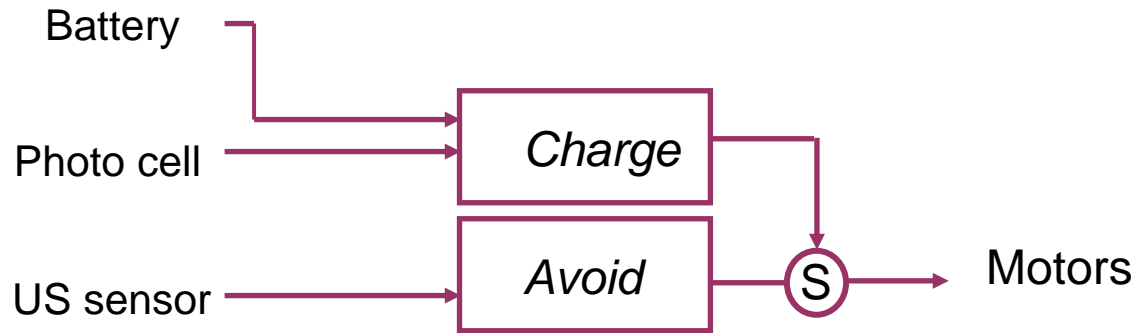
Note: responses are  
computed continuously  
(don't wait until finished)





# Subsumption Architecture

## Coordination of Behaviors

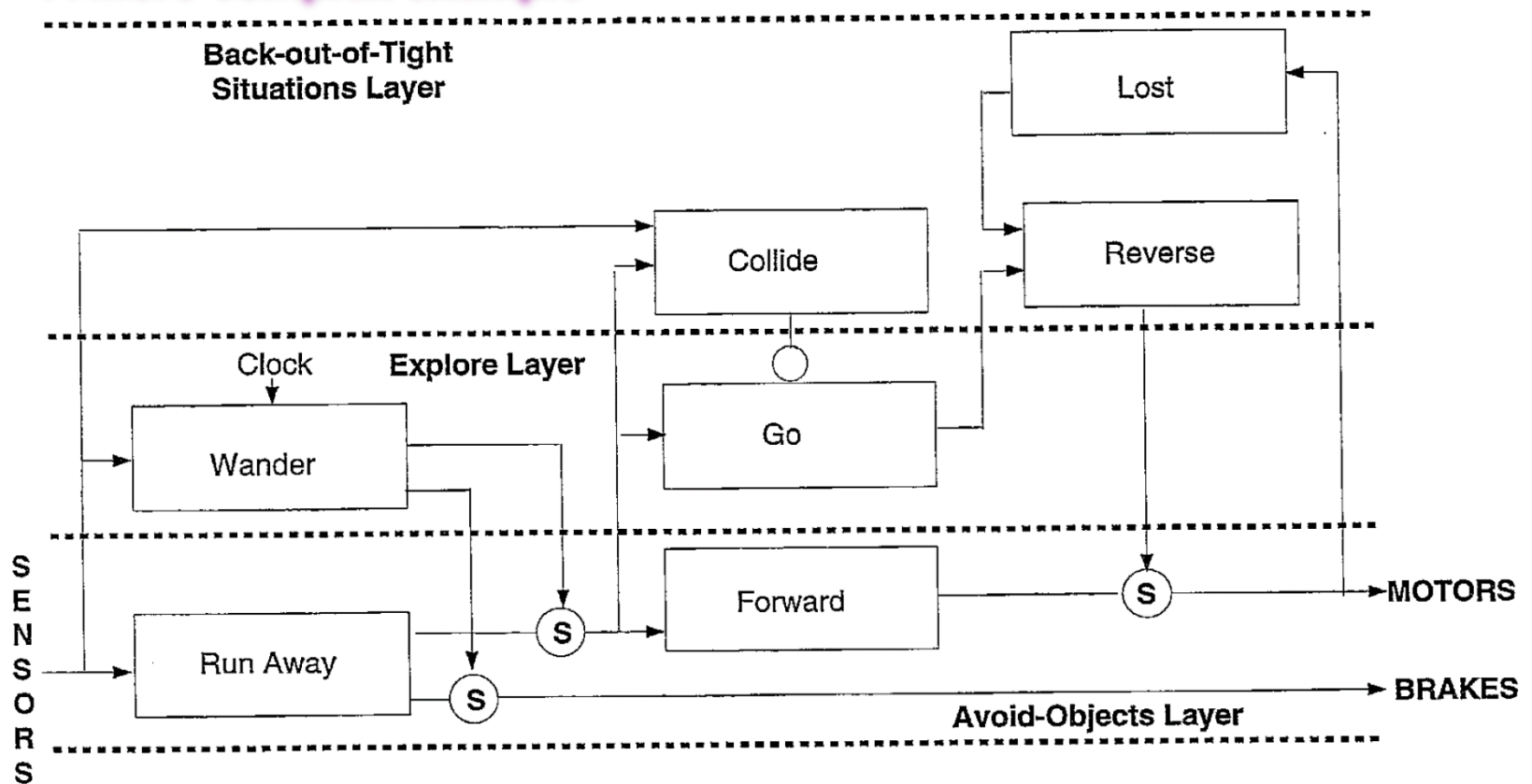


- **(S)** is a *suppressor node* that blocks and replaces messages from the *Avoid* behavior iff *Charge* sends messages at the same time.
- *Charge* subsumes (contains/innerluter) *Avoid*
- *Avoid* is subsumed by *Charge*



# Subsumption Architecture

## A more complex example



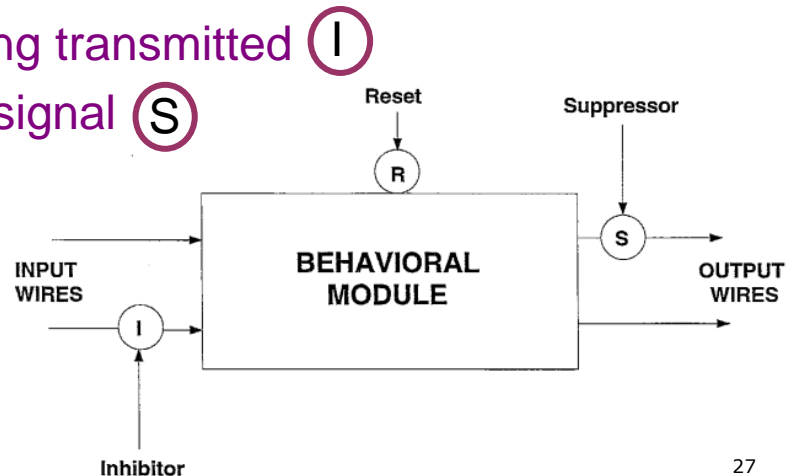
Each layer has a separate goal



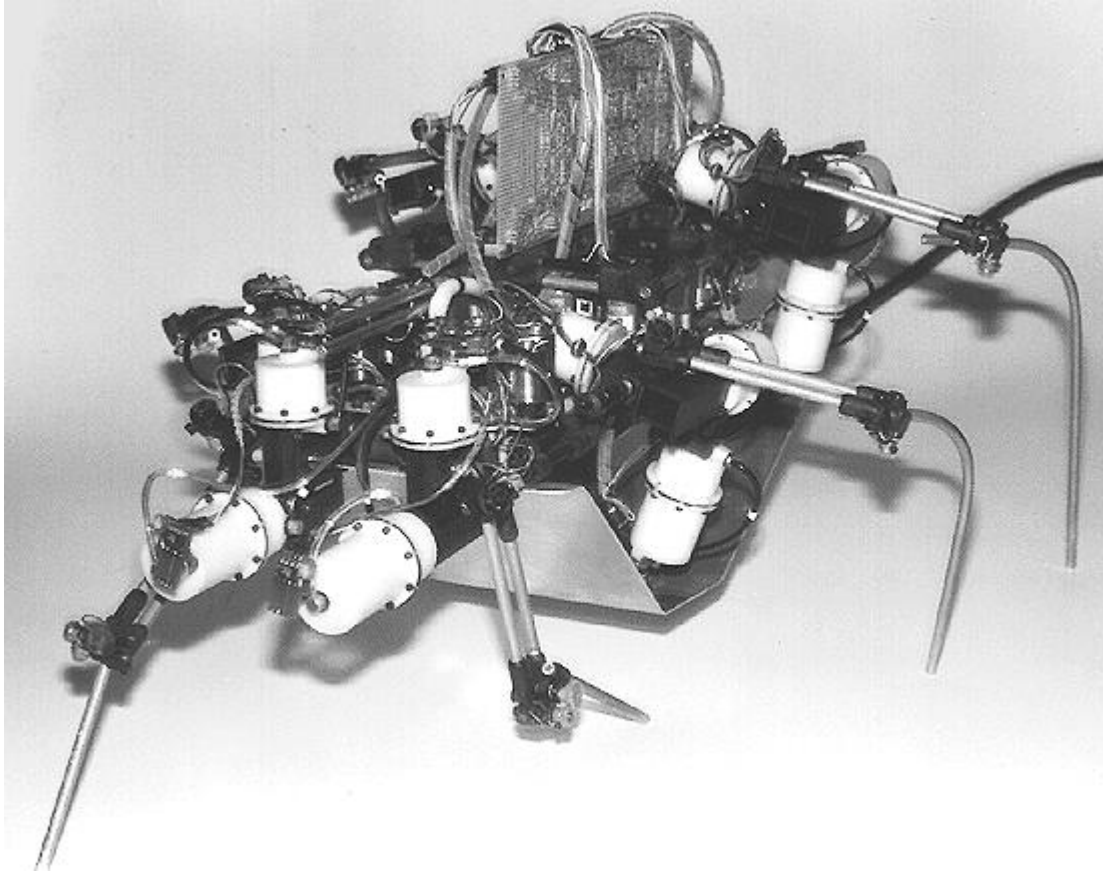
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# Subsumption Architecture

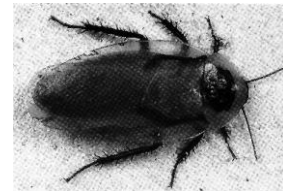
- Each task-achieving layer contains one or several reactive behaviors
- (almost) no internal states
- Loosely coupled layers:
  - The lower layers have no awareness of higher levels (i.e.: bottom-up design)
  - Higher layers may access lower levels, but very low bandwidth. “The world serves as primary medium of communication”
- Competitive coordination with two mechanisms:
  - Inhibition. Prevents a signal from being transmitted (I)
  - Suppression. Blocks and replaces a signal (S)

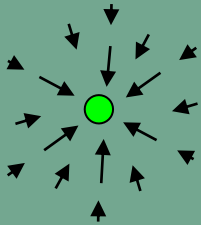
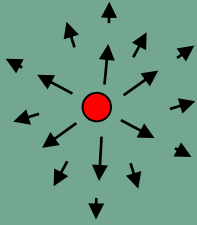


# A robot cockroach



- A six-legged, pneumatically powered walking robot based on the Subsumption architecture with 57 behaviors
- The robot's mechanics are modeled after a cockroach, *Blaberus Discoidalis*.





# POTENTIAL FIELDS FOR ROBOT NAVIGATION



## Robot Navigation with Potential Fields



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Dec 19, 2011  
UMINF-11.18  
ISSN-0348-0542



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# POTENTIAL FIELDS

- A reactive technique for guiding a robot from A to B
  - No Planning!
- Inspired by biology
- Responses are represented in a uniform format:  
Vectors with Strength (force or velocity) and Orientation
- Can be used to implement Motor schemas



# INSPIRED BY BIOLOGY



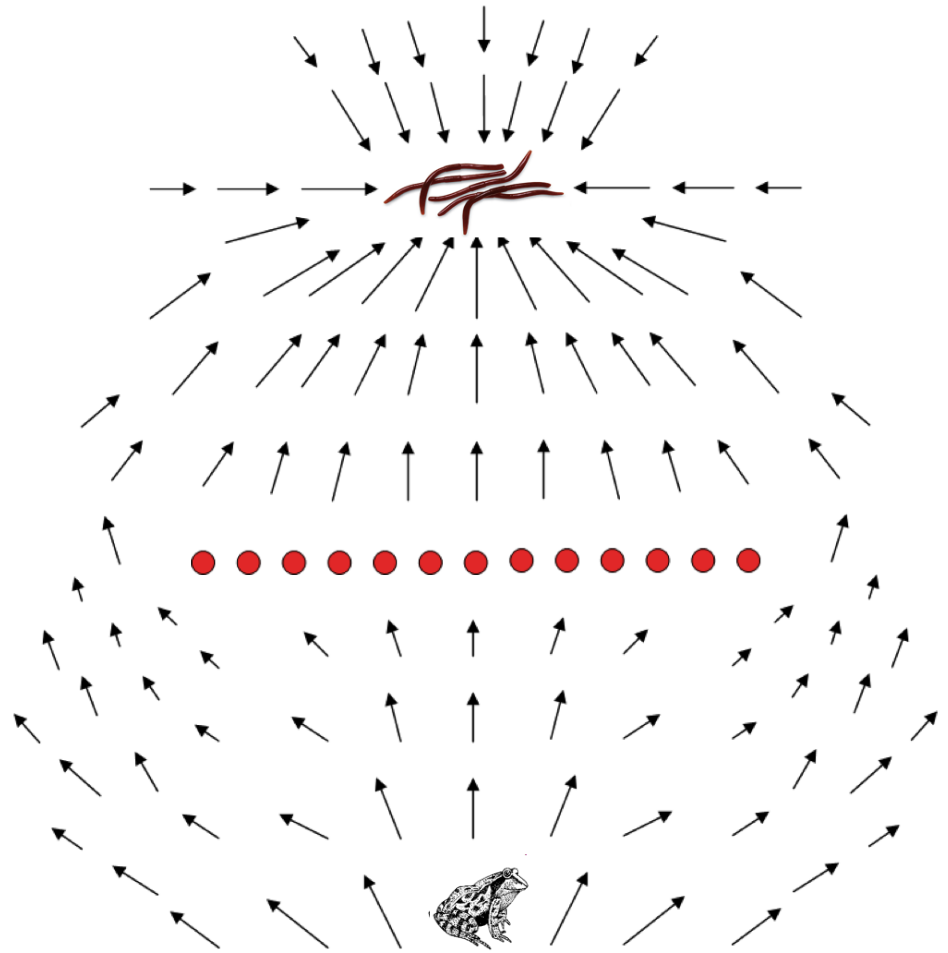
Describe the “forces” emanating from the shark and men below





# INSPIRED BY BIOLOGY

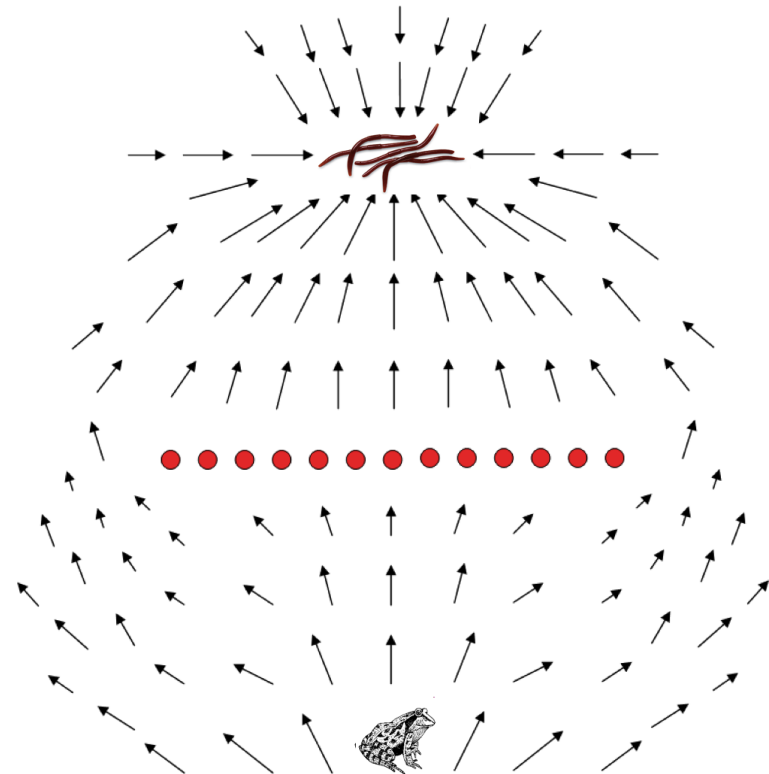
- Motion of toads attracted to a can of worms
- Model: Attraction and repulsion





# THEORY

- A robot should navigate over a 2D field occupied by an obstacle, until it reaches a goal position G.
- For each point (x,y) we define two values given by *potential field functions*, e.g.:
  - $U_{rep}$ :  $1/(\text{distance to obstacle})$
  - $U_{att}$ : distance to G
- We want to move in the direction of decreasing  $U_{rep}$  and  $U_{att}$



# THEORY

- For a function  $f$ , the gradient vector  $\nabla f$  points in the direction of increasing  $f$ .
- We define *force vectors*  $F$  as the negative gradients of  $U_{rep}$  and  $U_{att}$ :

$$F_{rep} = -\nabla U_{rep} = -\left(\frac{\partial U_{rep}}{\partial x}, \frac{\partial U_{rep}}{\partial y}\right)$$
$$F_{att} = -\nabla U_{att} = -\left(\frac{\partial U_{att}}{\partial x}, \frac{\partial U_{att}}{\partial y}\right)$$

- $F_{rep}$  points in the directions of smaller  $U_{rep}$  (away from the obstacle)
- $F_{att}$  points in the directions of smaller  $U_{att}$  (towards G)
- A strategy to move safely towards G is to move in the direction  $F = F_{rep} + F_{att}$



# THEORY

## Generalization:

- More obstacles
- Other types of potential fields

The general expression for  $\mathbf{F}$  computed at a point  $(x,y)$  is:

$$\mathbf{F}(x, y) = \sum_i \mathbf{F}_i(x, y)$$

where  $\mathbf{F}_i(x,y)$  is the force vector associated with potential field  $U_i$ , i.e. the negative gradient of  $U_i$  :  $\mathbf{F}_i(x,y) = -\nabla U_i(x,y)$



# AN ALGORITHM FOR NAVIGATION

An algorithm for robot navigation from a position  $(x,y)$ :

1. Represent the goal and all obstacles as potential fields.
2. Compute force vectors  $\mathbf{F}_i(x,y)$  for all potential fields
3. Sum all force vectors into  $\mathbf{F}(x,y)$
4. Move along  $\mathbf{F}(x,y)$  with speed proportional to  $|\mathbf{F}(x,y)|$
5. Repeat from 2. until goal is reached

## IMPORTANT:

- Normally each  $\mathbf{F}_i$  is computed as a function of distance and direction to objects around the robot (from sensors)
- $\mathbf{F}_i$  only has to be computed for current position  $(x,y)$



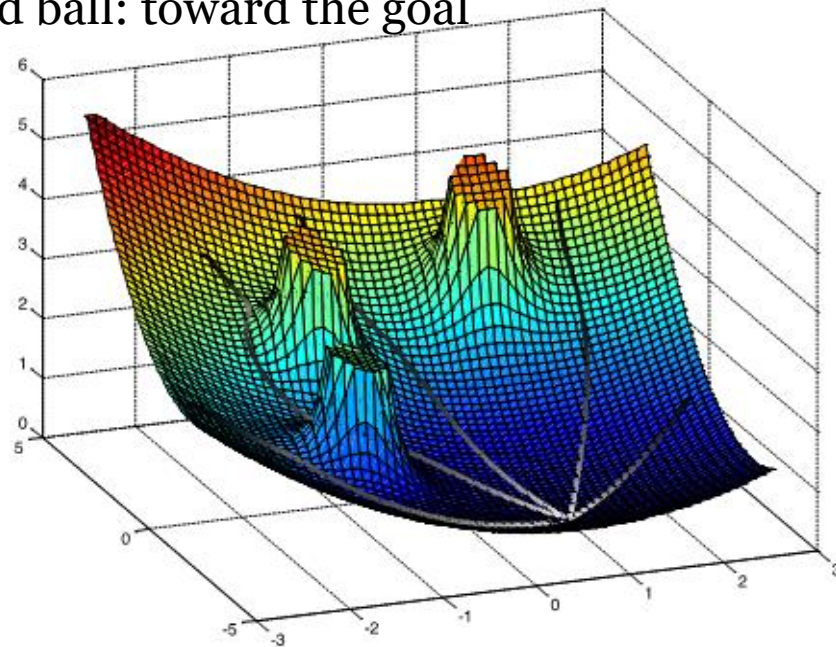
# PHYSICS ANALOGY

A ball finding its way to the bottom of a valley filled with obstacles

- One attractive goal at (1, -2) at the bottom of the valley
- Three repulsive objects as “steep rocks”
- The paths show the motion of the imagined ball: toward the goal avoiding the obstacles

The paths also corresponds to the potential field theory:

- z-axis is the sum  $U$  of all potential fields
- Direction of maximum descent ( $-\nabla U$ )

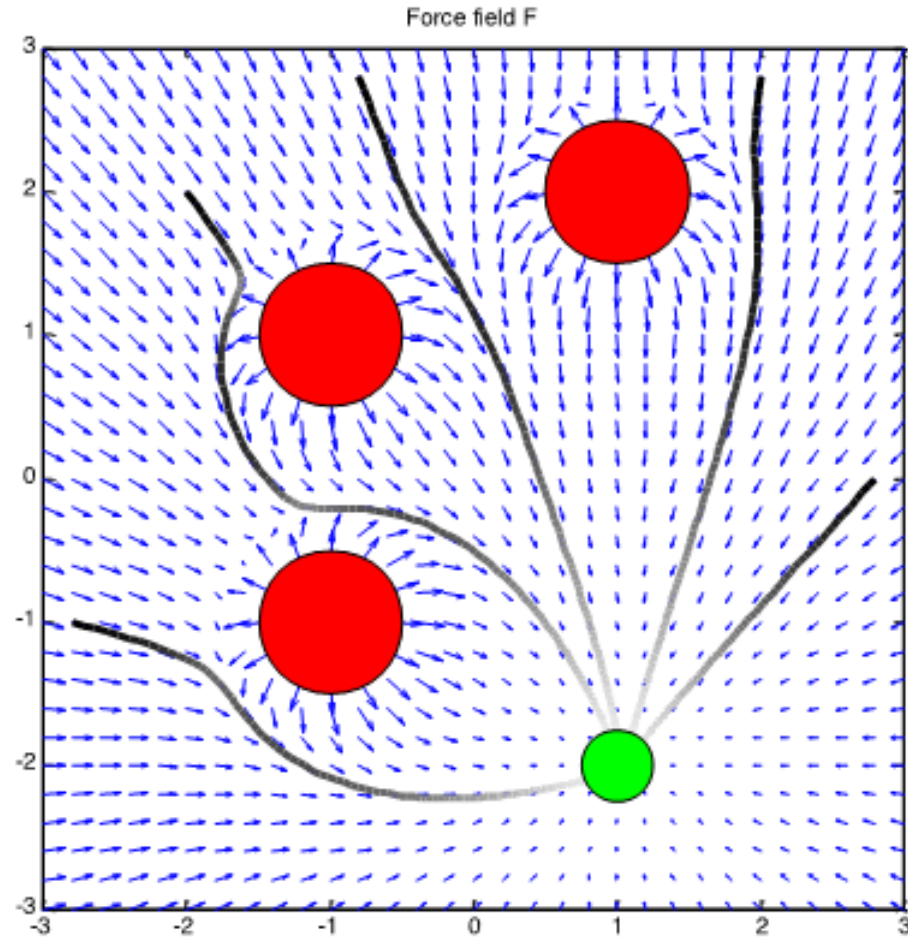


# PHYSICS ANALOGY CONT.

- $\mathbf{F} = -\nabla U$  is the force that causes the ball to move
- Each arrow represents the direction and magnitude of  $\mathbf{F}$  at the location of the arrow
- The robot follows the direction of  $\mathbf{F}$  in every point
- Speed is given by  $|\mathbf{F}|$ , the magnitude of  $\mathbf{F}$  (shown as gray scale of paths)
- Compare with the toad and worms

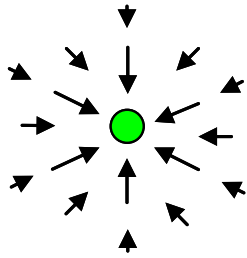
Red: repulsive obstacles

Green: attractive goal

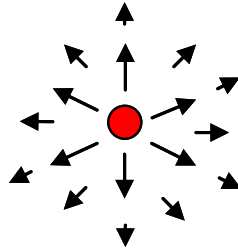


# BASIC TYPES OF POTENTIAL FIELDS

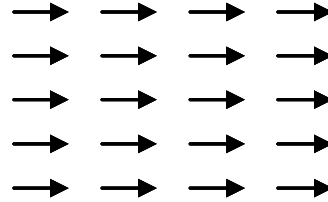
Attraction



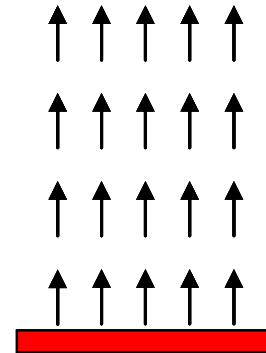
Repulsion



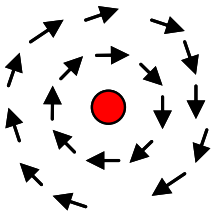
Uniform



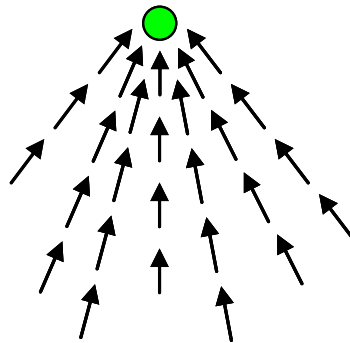
Perpendicular



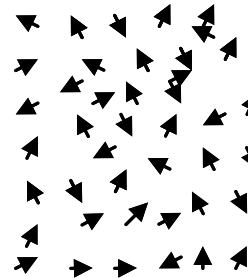
Tangential



Selective attraction



Random



- The figures show the force field  $\mathbf{F}$ , i.e. the gradient of the potential field function
- Each arrow indicates strength and direction of the force
- The force fields are often, somewhat sloppily, denoted potential fields

# FIELDS AS BEHAVIORS

Basic idea: Follow the force!

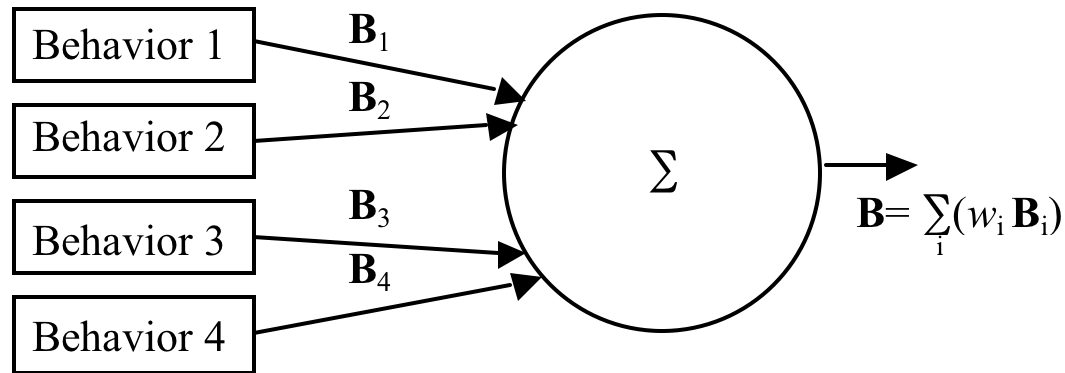
- One field can be thought of as one behavior
- Vector summation to create compound behaviors



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# FIELDS AS BEHAVIORS



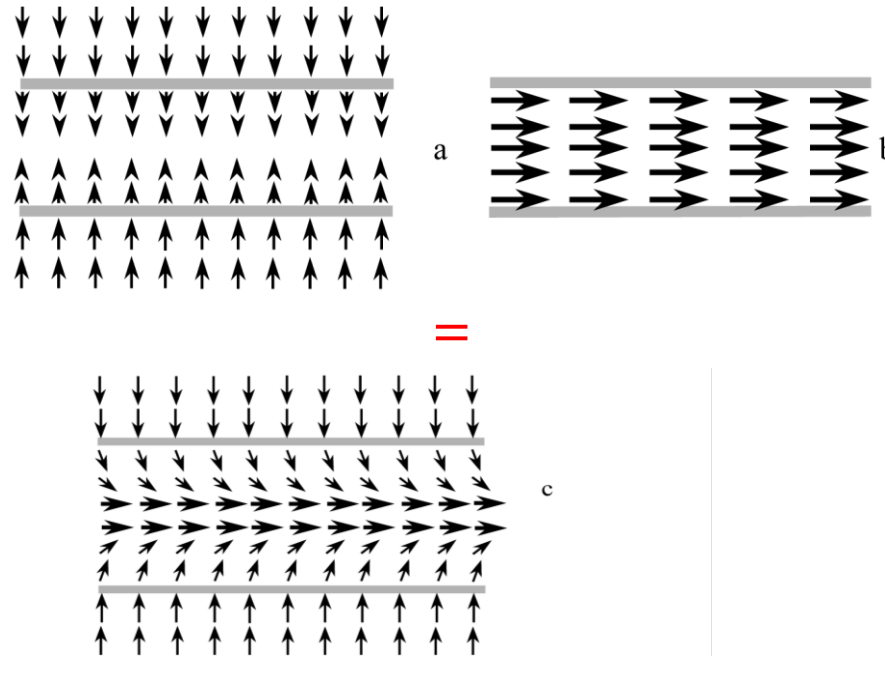
- Four behaviors; each one represented by a potential field, or recursively by other coordinated behaviors
- The output force vectors  $\mathbf{B}_1, \dots, \mathbf{B}_4$  are summed and weighted to produce one compound vector output  $\mathbf{B}$
- Note:  $\mathbf{B}$  needs to be computed for the robot's location only



# EXAMPLE: FOLLOW-CORRIDOR

constructed as a sum of 3 potential fields:

2 x Perpendicular + Uniform



The robot “feels” the vector for the current location, then moves accordingly, “feels” the next vector, moves, etc.

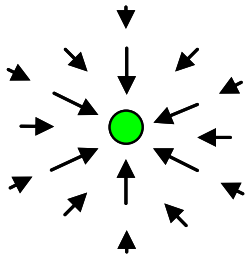


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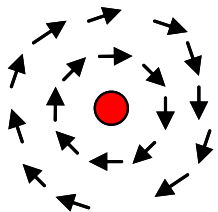
# EXAMPLE: DOCKING BEHAVIOR

- Constructed as a sum of 3 potential fields:

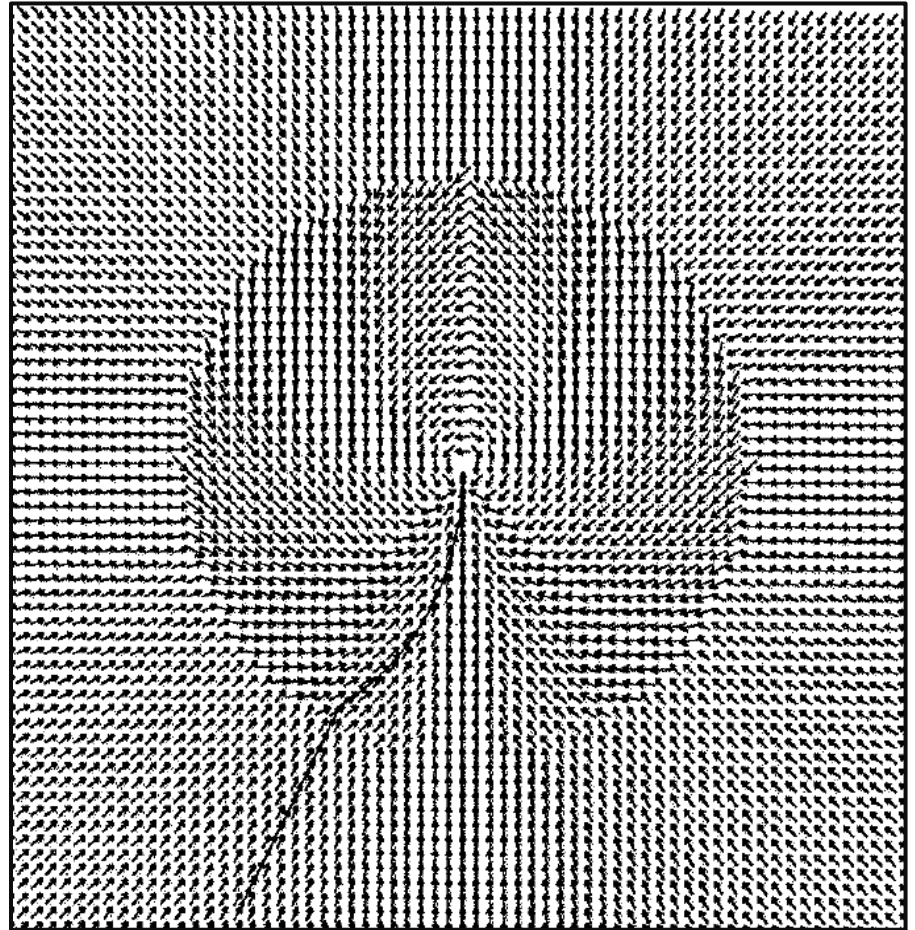
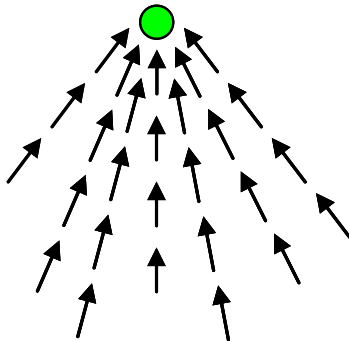
Attraction



Tangential



Selective attraction



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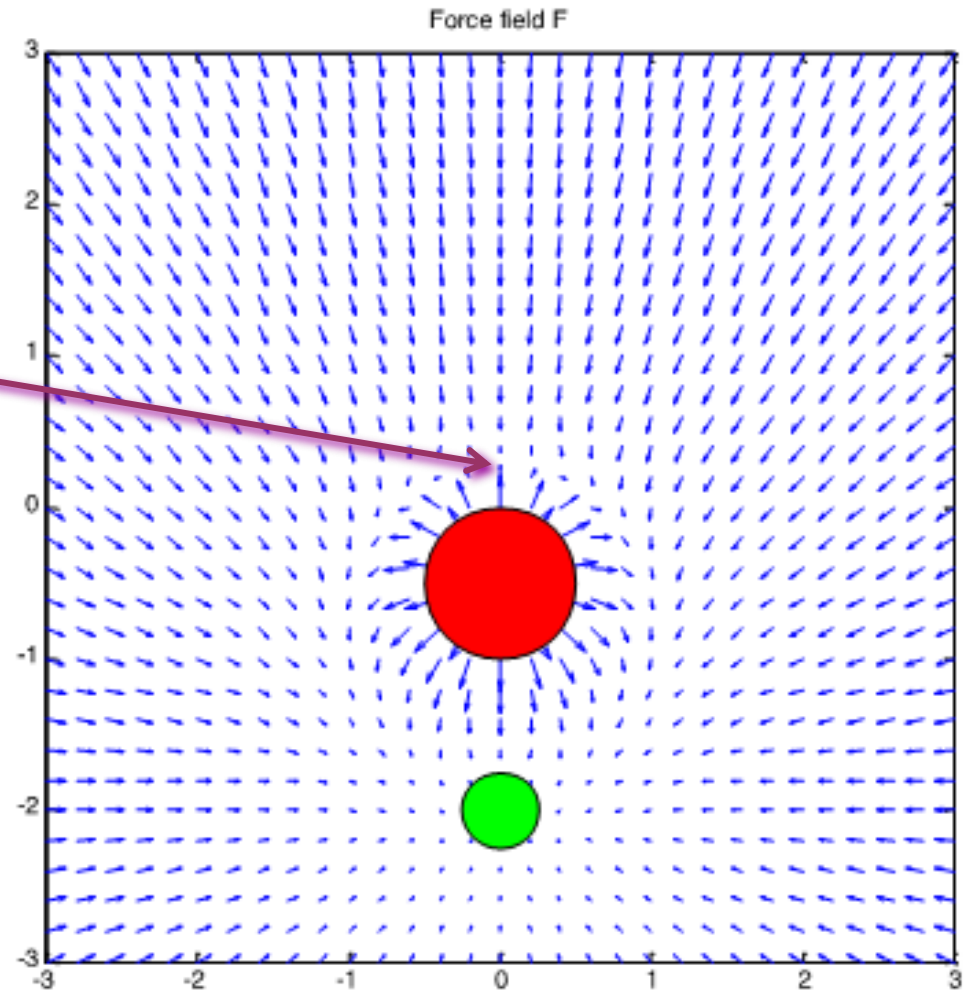
# PROBLEM: LOCAL MINIMA

Sometimes the fields add to zero at some points!

## Example:

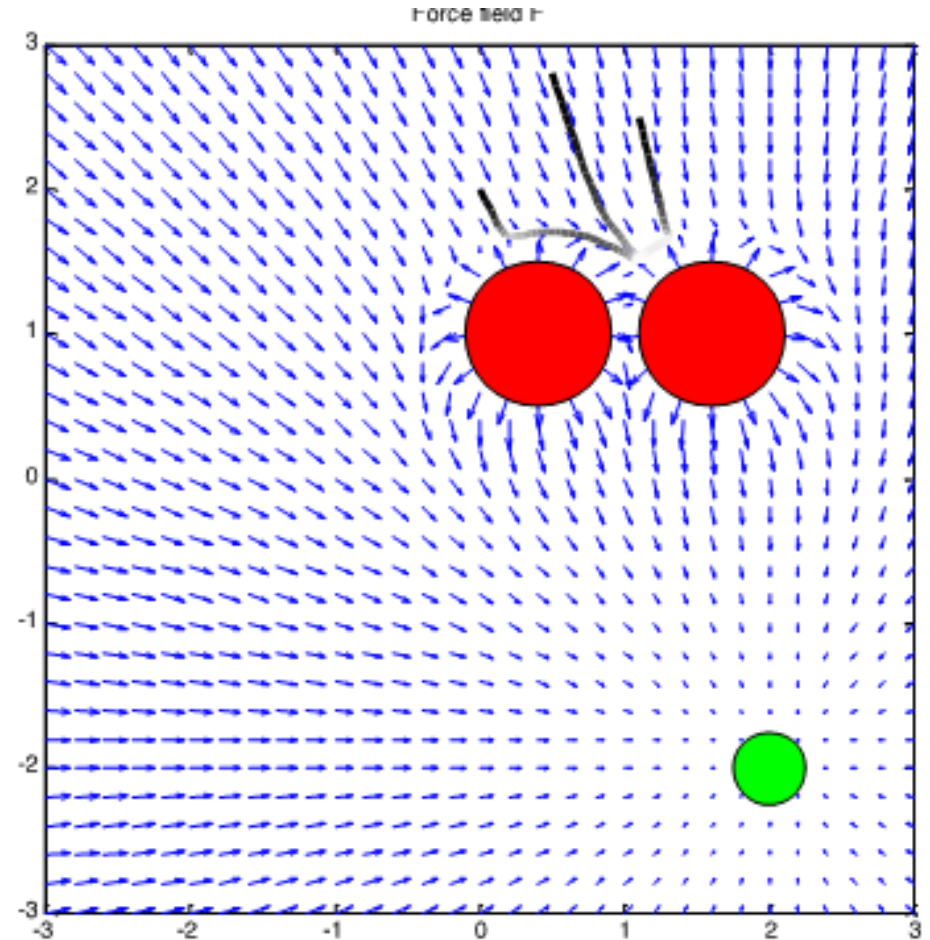
The repulsive force from the obstacle cancels the attractive force from the goal.

If the robot reaches this point, it will stop!



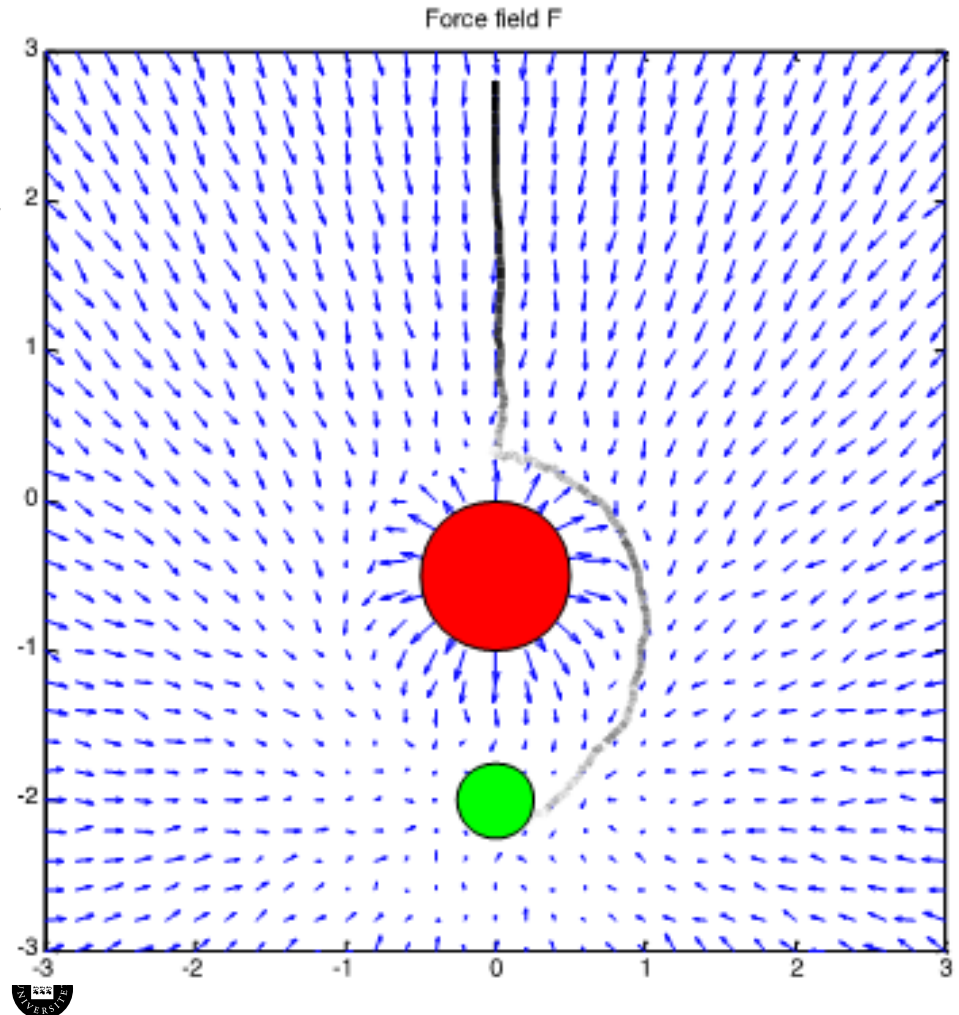
# PROBLEM: LOCAL MINIMA

- The robot may be dragged into a local minimum from several different starting positions.
- The gray scale of robot paths corresponds to the speed of the robot.



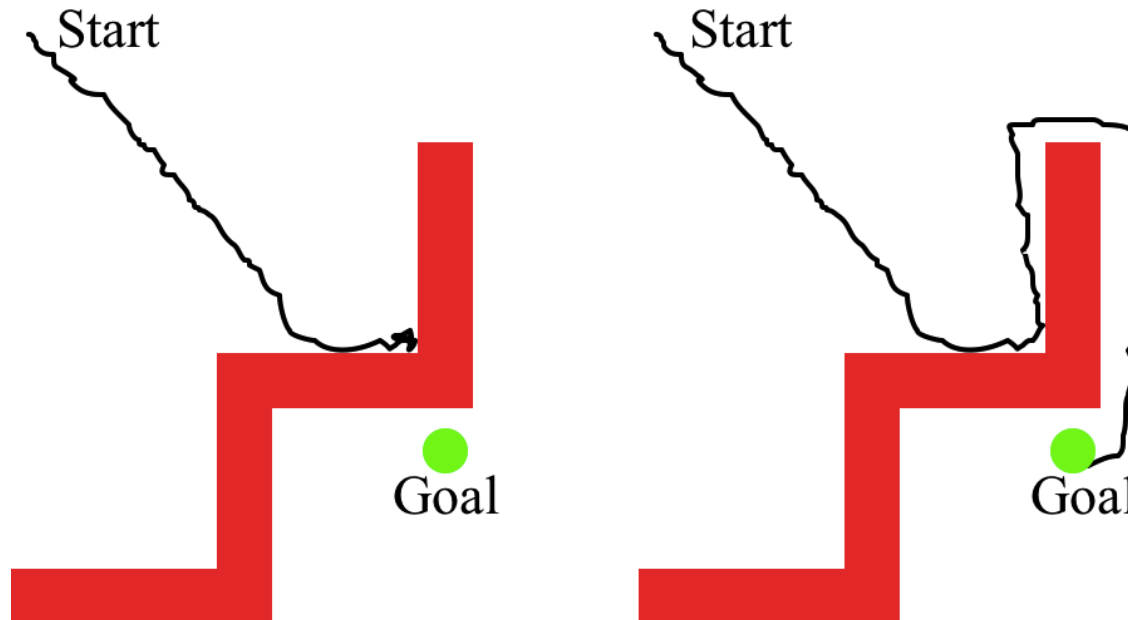
# SOLUTION 1: LOCAL MINIMA

- The local minimum is avoided by adding a potential field comprising random values.
- The jagged path is a result of the random force vectors.



## SOLUTION 2: LOCAL MINIMA

- Sometimes, adding noise will not solve the problem but rather introduce a **cyclic behavior**: the robot keeps returning to the local minimum
- Solution: Adding a dynamically updated potential field with repulsive forces from all previously visited locations

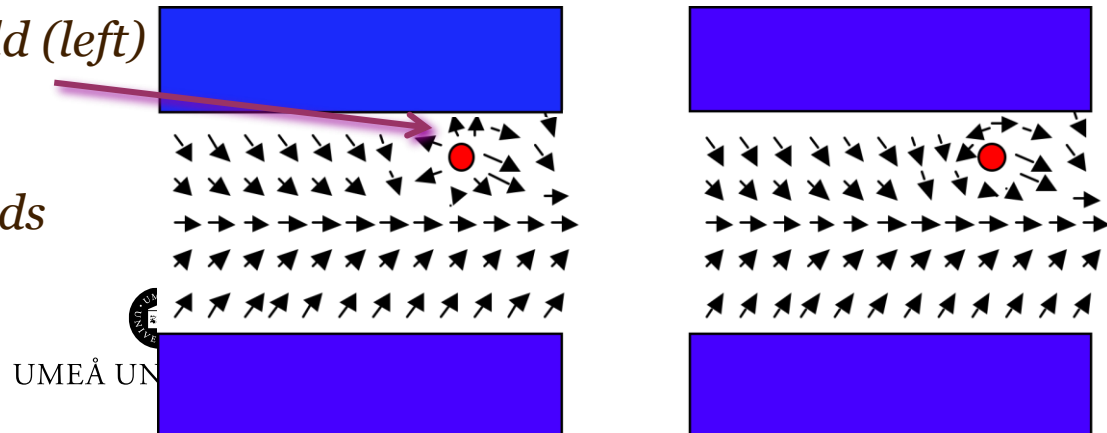


# SOLUTION 3: LOCAL MINIMA

- The regular repulsive fields are replaced by a special kind of tangential fields : Navigation templates
- The direction of the tangential field is set in real-time not to counter the vector sum of all other fields
- Reduces the risk of attractive and repulsive forces cancelling each other
- Another benefit: Obstacles will be avoided by a force vector that points in the same direction as the general behavior:

*The regular repulsive potential field (left) may push the robot into the water*

*A navigation template (right) avoids this by forcing the robot to avoid the obstacle in a safe direction*





# PROBLEMS WITH POTENTIAL FIELDS

- Jerky motion (depends on update rates)
- Robot and obstacles treated as points ( $\Rightarrow$  risk of collision)
- The robot is expected to change velocity and direction instantaneously
- Sensitive to local minima

Three approaches:

- Injecting randomness
- Adding an “avoid-past” schema
- Navigation templates

- Cyclic behavior

One approach:

- Adding an “avoid-past” schema that generates a repulsive force from recently visited areas



# ADVANTAGES WITH POTENTIAL FIELDS

- Easy to implement and visualize
  - Robot behavior easy to predict by designer
- Support for parallelism
  - Each field is independent of the others and may be implemented as general software, or even hardware, modules
- They can be easily parameterized and configured, also during design or in real-time
- The combination mechanism is flexible and can be tweaked with gains to reflect varying importance of sub-behaviors



# **COORDINATION OF BEHAVIORS**

**A general problem in reactive systems  
is how to handle concurrent and  
disagreeing behaviors**



# COORDINATION OF BEHAVIORS

Two major types of coordination mechanisms:

**Competitive.** “Winner takes all”, “Arbitration”

- Fixed priority methods
  - Subsumption
  - ”Subsumption light”
- Dynamic priority methods
  - Activity based; Select the most “active” behavior.
  - Voting based; each behavior suggests MANY responses

**Cooperative.** “Behavioral fusion”

- The representation of behaviors must allow fusion.
  - Vector addition such as in potential-fields and schema based approaches
  - Vector addition with dynamic gains
  - Fuzzy logic



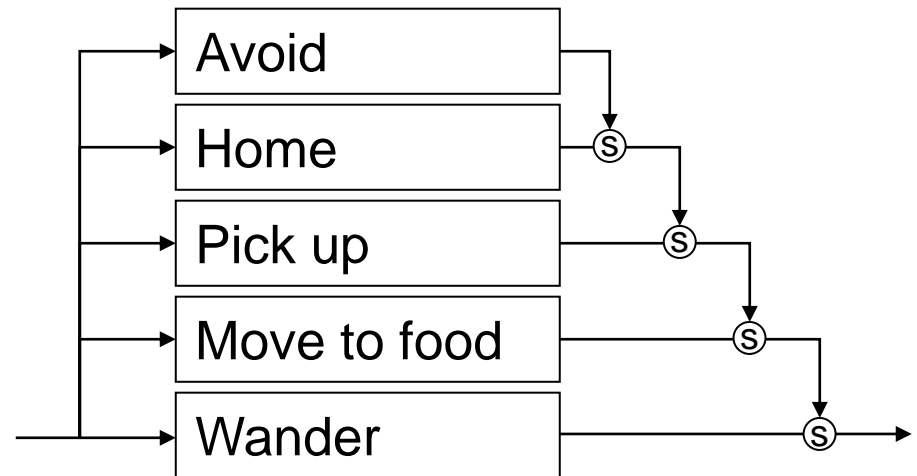
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# "SUBSUMPTION LIGHT"

Competitive - Fixed priority

## Foraging example:

- *Wander*: always active
- *Move to food*: active if the robot sees the food
- *Pick up*: active if food is within reach
- *Home*: active if the robot has food
- *Avoid*: active when an obstacle is in front of the robot



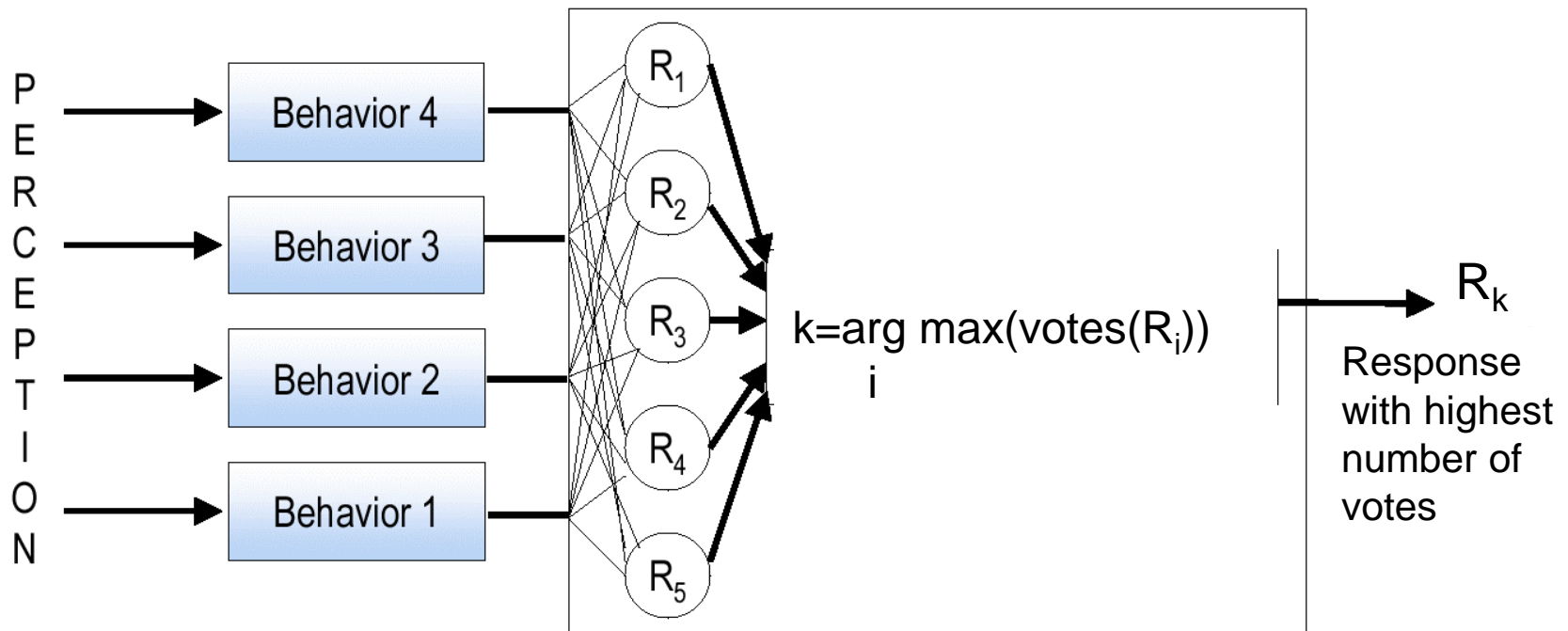
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# Voting based arbitration

## Competitive - Dynamic priority

- One pre-defined set of discrete Responses for ALL behaviors
- Each behavior votes for each response
- The response with most votes is executed

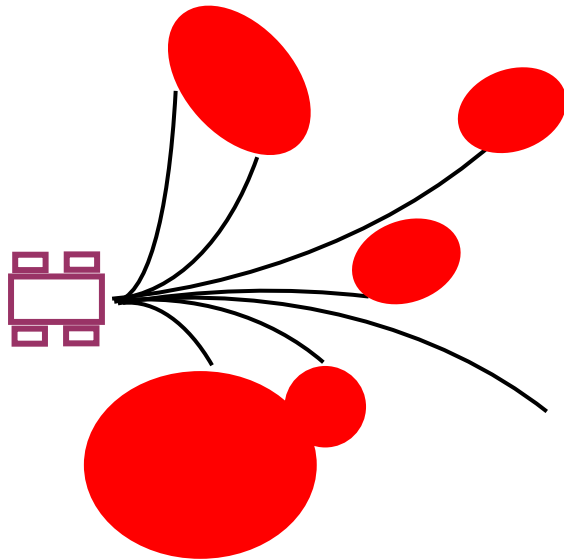
Priorities (no. of votes) may vary during mission



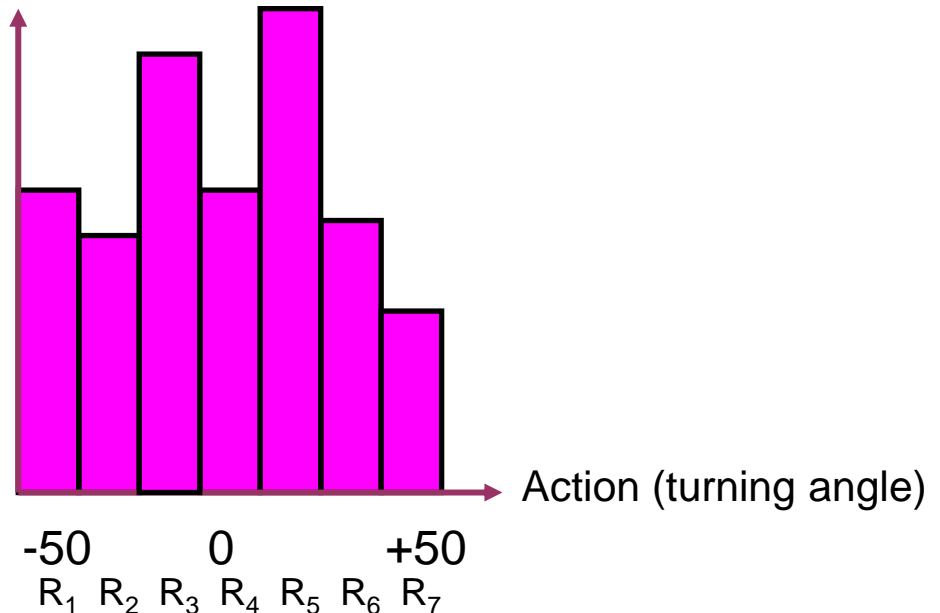
Voting-based coordination

# EXAMPLE VOTING

- DAMN – A Distributed Architecture for Mobile Navigation
- Behavior 1 does obstacle avoidance and votes for each response based on estimated time to collision in that direction.
- Behavior 2 does homing and votes for each response based on direction to goal
- Responses R1 to R7 : turning angle



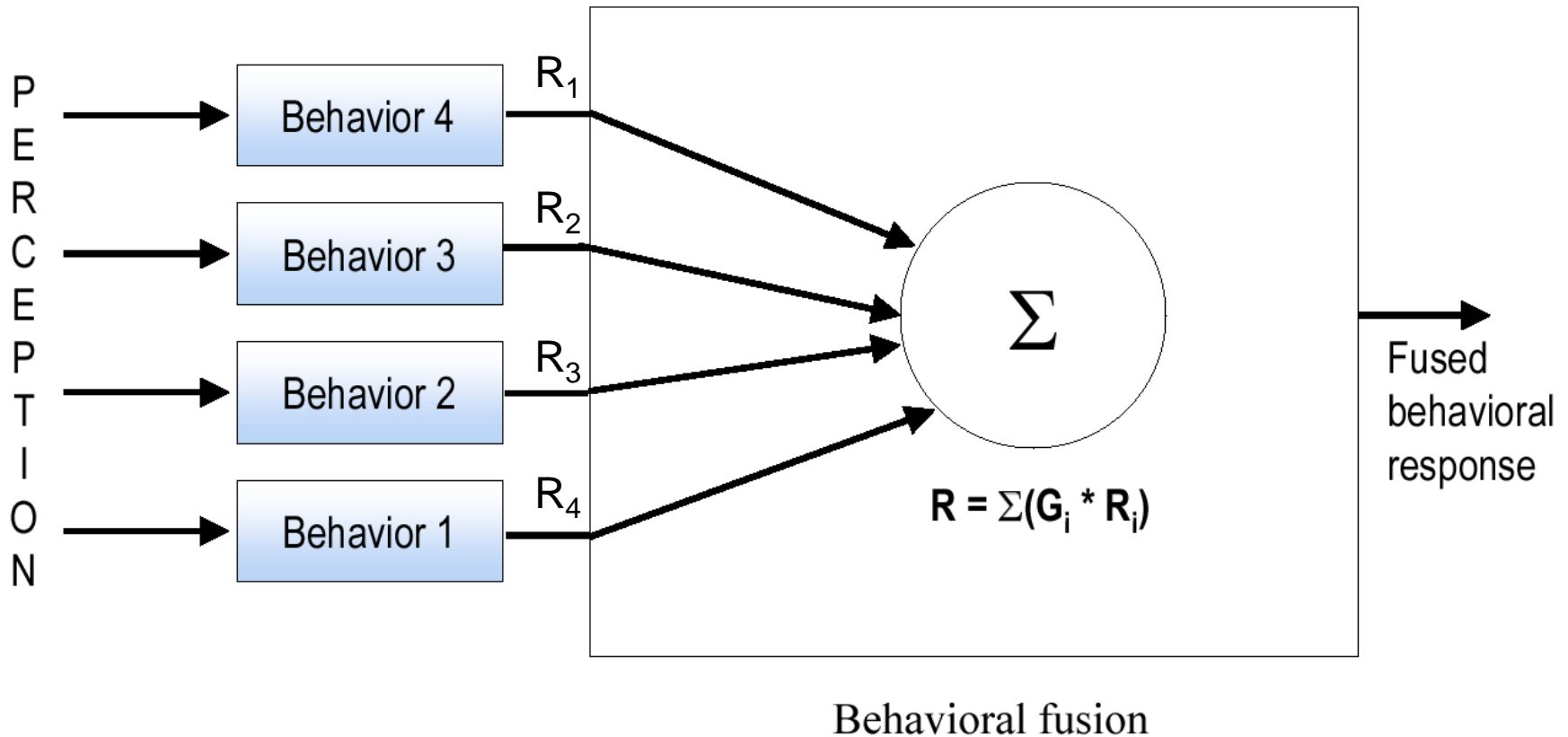
Votes  
by Behavior 1



# Vector Addition

Cooperative – dynamic gains

The behaviors are fused using gains  $G_i$ . Gain levels may vary during mission and depend on the robot's goals and percepts



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# THE REACTIVE PARADIGM

## Advantages:

- Fast reaction time
- Requires less memory and CPU power
- Easy to implement and expand
  - Behaviors are independent
  - No modelling necessary
- Makes no Closed World Assumption
  - Works in dynamic environments that are difficult to characterize and contain a lot of uncertainty



# THE REACTIVE PARADIGM

## Disadvantages:

- Interacting behaviors can be unpredictable  
*“fast, cheap and out of control...”* (Rodney Brooks)
- Low level intelligence
  - No memory:
    - Can't handle sequences of behaviors
    - Can't learn anything
  - No world representations:
    - Doesn't know where it is
- Often more art than science



# **HYBRID DELIBERATIVE/REACTIVE PARADIGM**

- Reactivity by the end of the 1980's was the major trend!
- But not remembering the state of the robot and the world is limiting!
- Question:
  - Should planning be reintroduced? How?
- Answer:
  - Hybrid systems!
  - More on this in the next AI course!





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