64-424
Intelligente Roboter

http://tams.informatik.uni-hamburg.de/lectures/2011ws/vorlesung/ir

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Fakultät für Mathematik, Informatik und Naturwissenschaften
Fachbereich Informatik

Technische Aspekte Multimodaler Systeme

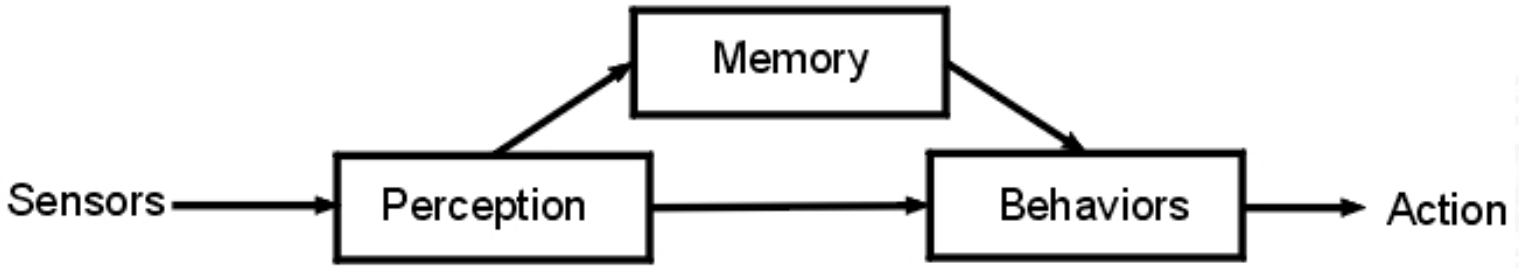
Wintersemester 2011/2012

Gliederung

Gliederung (cont.)

Interaktive Architekturen

Das Perzeption-Aktion-Modell mit Gedächtnis



Das CMAC-Modell

CMAC: „Cerebellar Model Articulation Controller“ (Englisch)

- The input vector **S** is modelled as sensory firing cell patterns.
- The combination of the cell patterns produces an association vector **A**. This association cell vector multiplied by a memory matrix W produces a response vector **P**.
- The CMAC model can be viewed as two mappings:

$$f : \mathbf{S} \rightarrow \mathbf{A}$$

$$g : \mathbf{A} \rightarrow \mathbf{P}$$

$$\mathbf{S} = \{\textit{sensory input vectors}\}$$

$$\mathbf{A} = \{\textit{association cell vectors}\}$$

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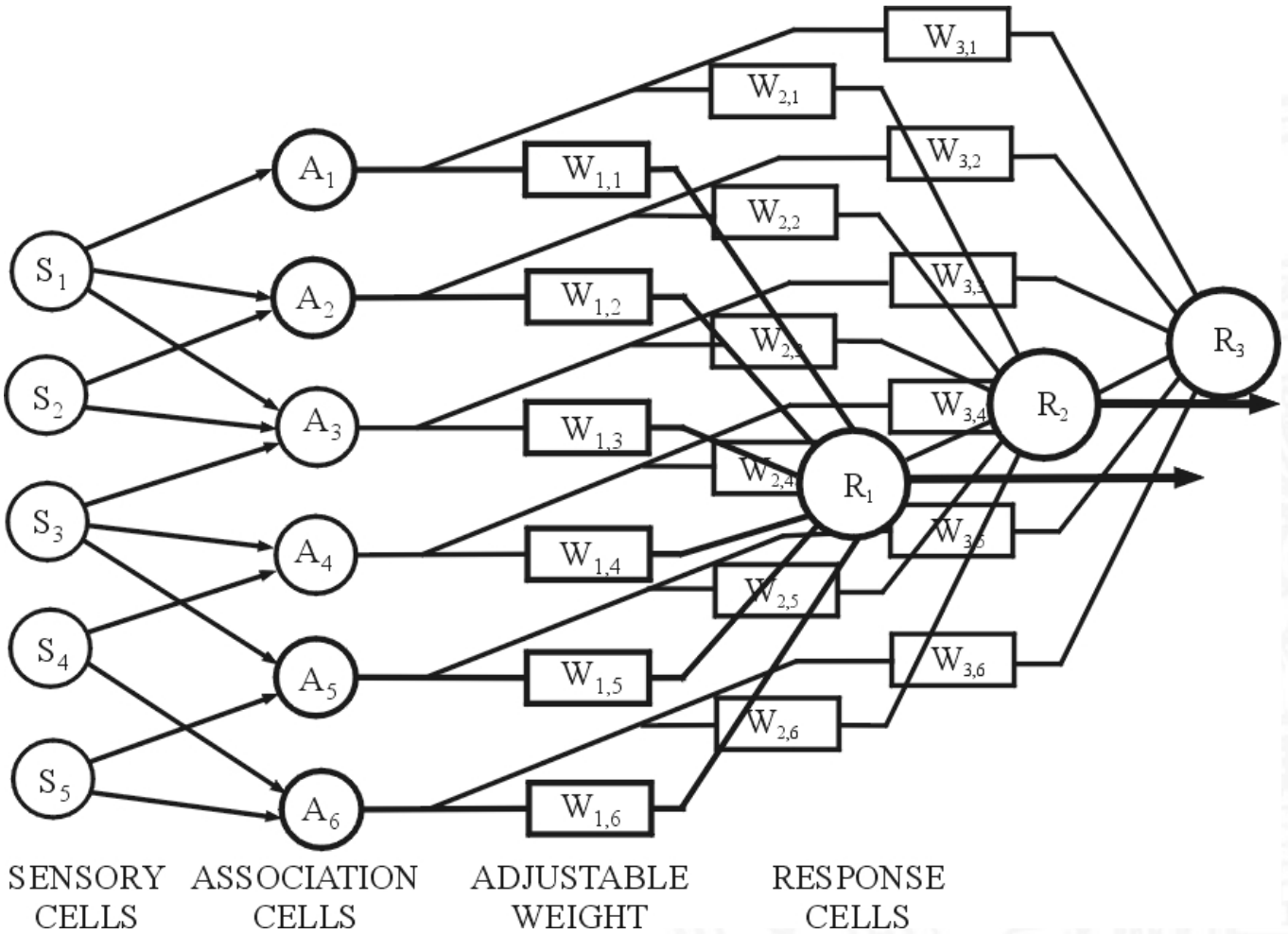

B-Spline-Modell

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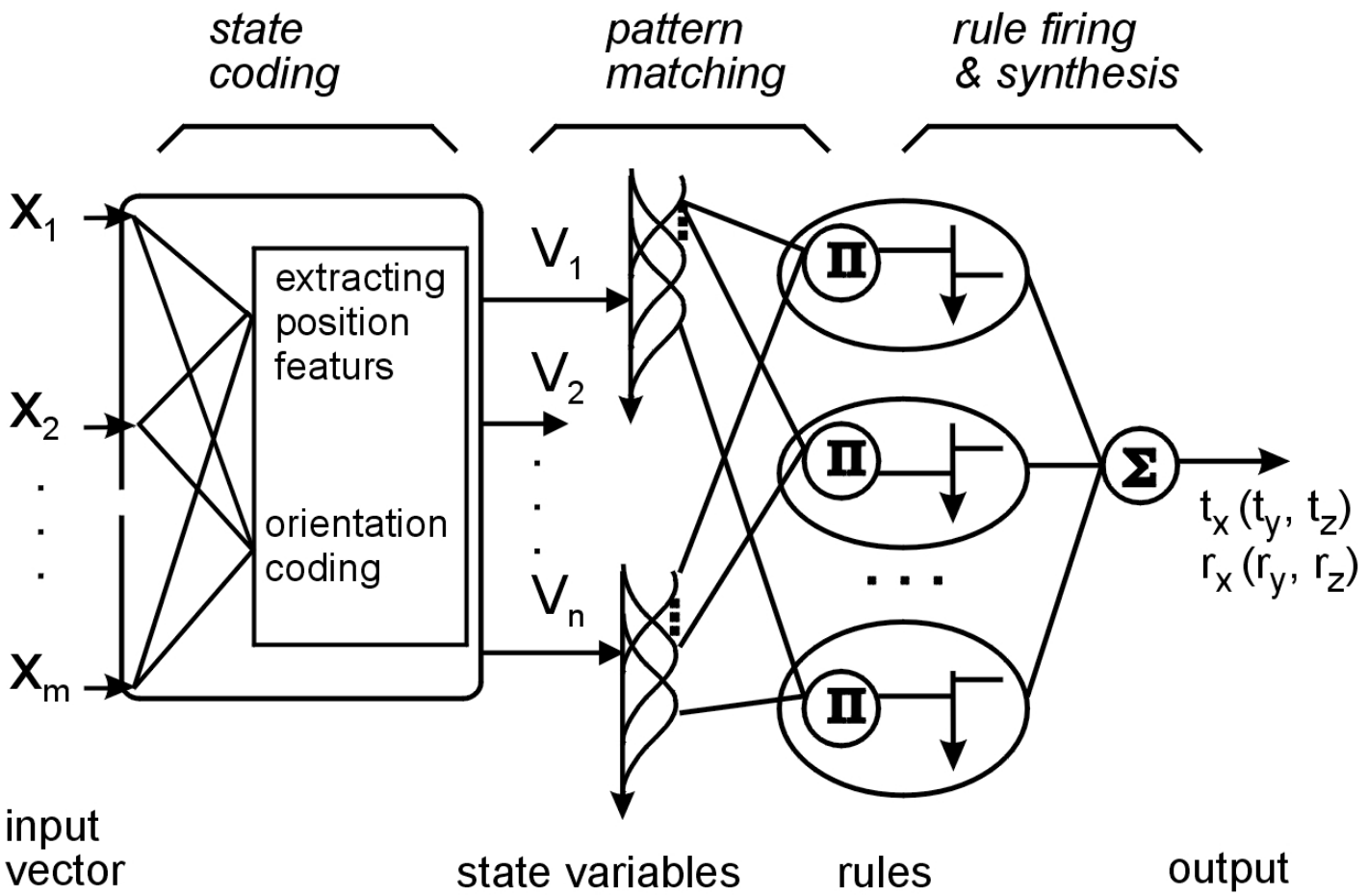
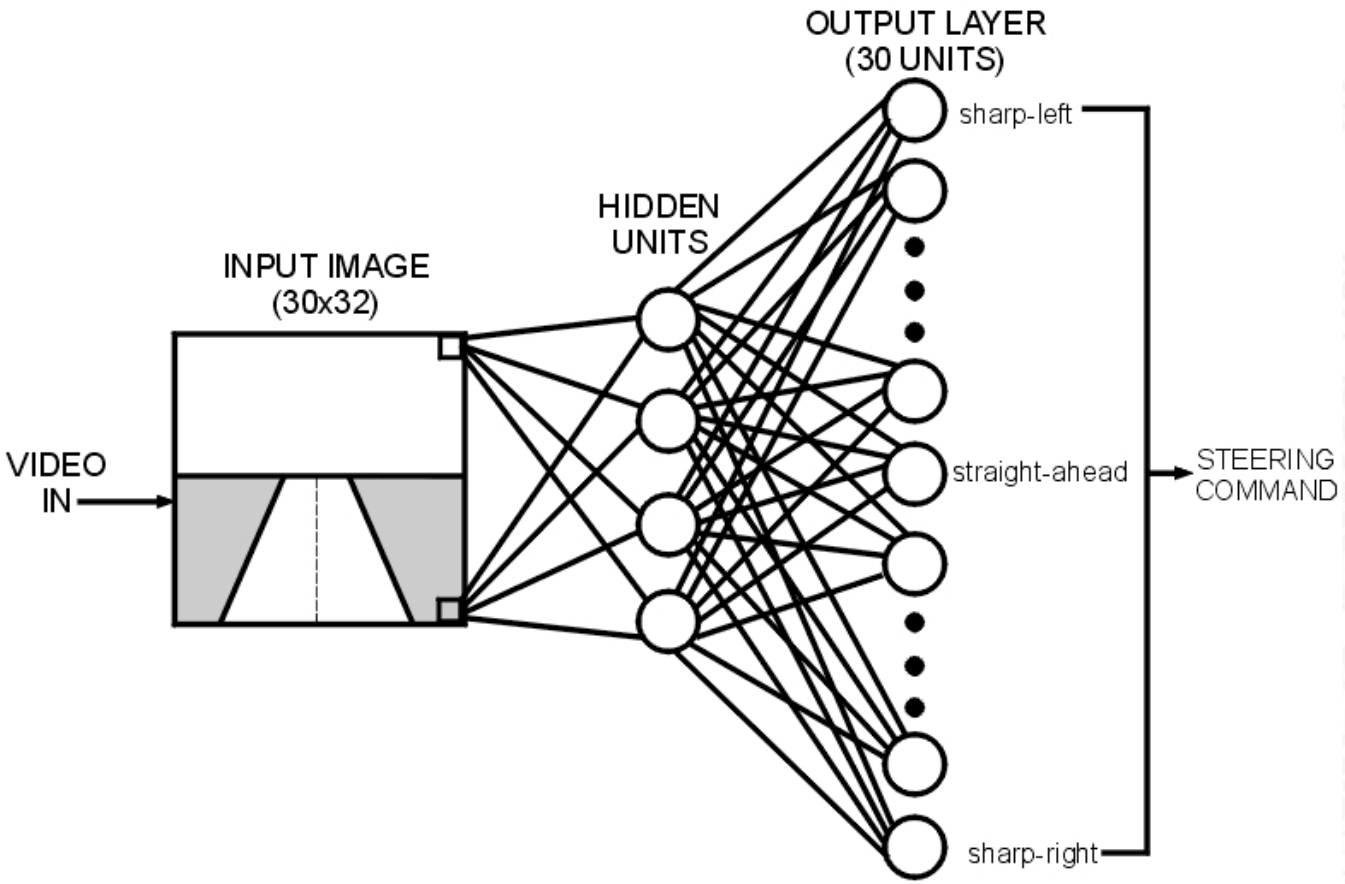
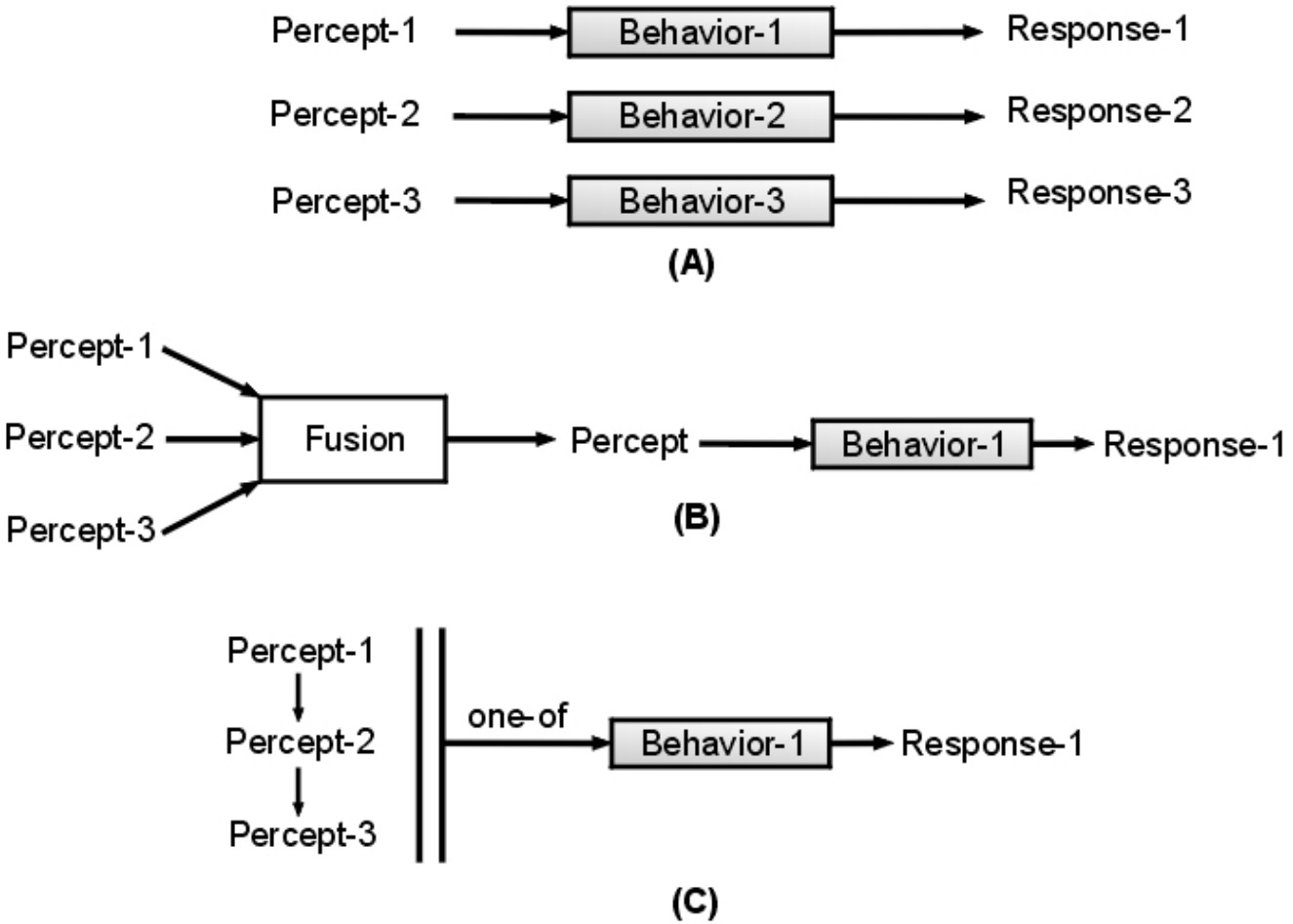


Alvinn - Visuelle Navigation CMU



Handlungsorientierte Perzeption



Die Subsumptions-Architektur

A robot program employing the modelling/planning paradigm is composed of a sequence of steps. These functional units transform a snapshot of sensory data into a series of actions inteded to achieve a specified goal. (Englisch, R. Brooks)

Die Subsumptions-Architektur (cont.)

- A subsumption architecture for a mobile robot „Rug Warrior“ begins with a behaviour called „Cruise“ (moving forward).
- „Follow“ is triggered by photo-cells to move toward right.
- „Avoid“ suppresses „Follow“ and „Cruise“ when the near-infrared sensors detect an imminent collision and „Escape“ also helps to avoid obstacles.
- The highest-level behaviour, „Detect-Sound-Pattern“ caused the robot to play a tune.

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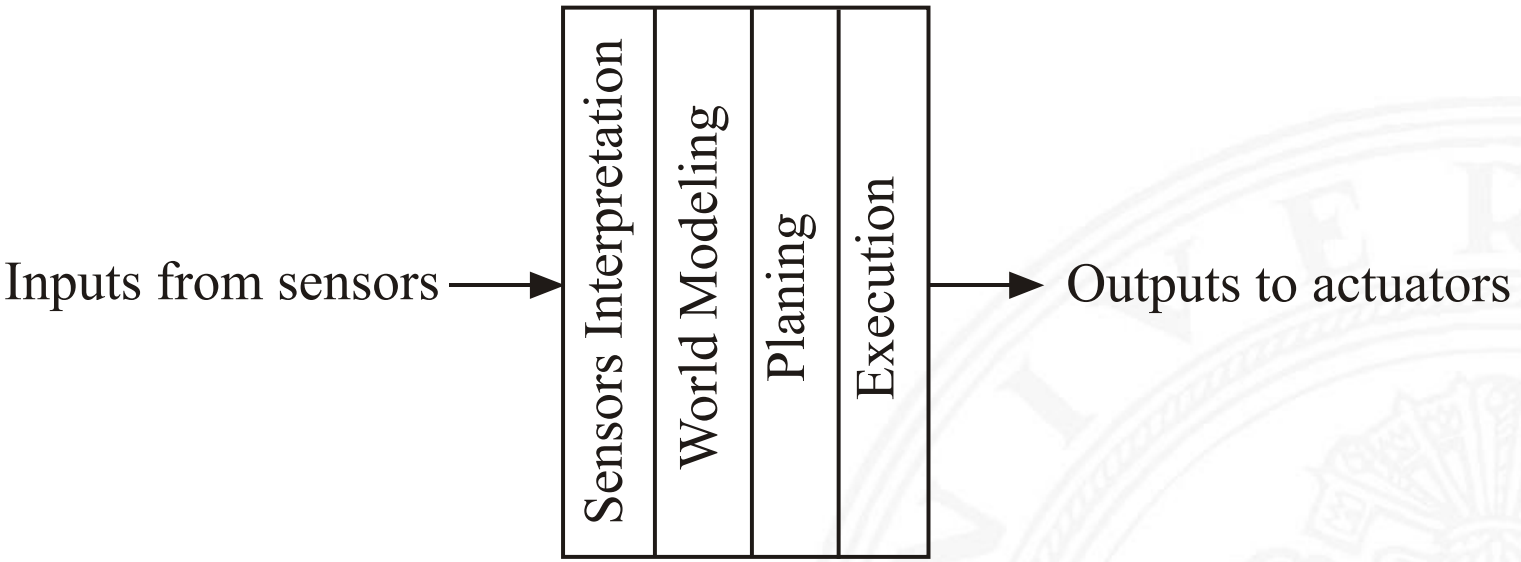

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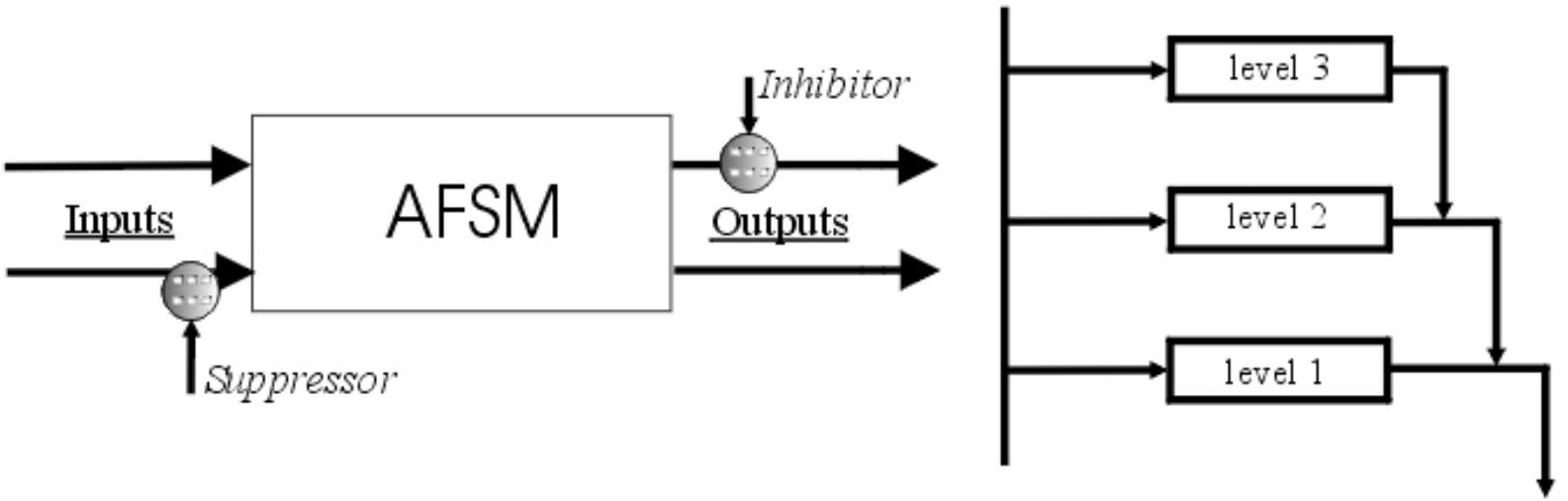

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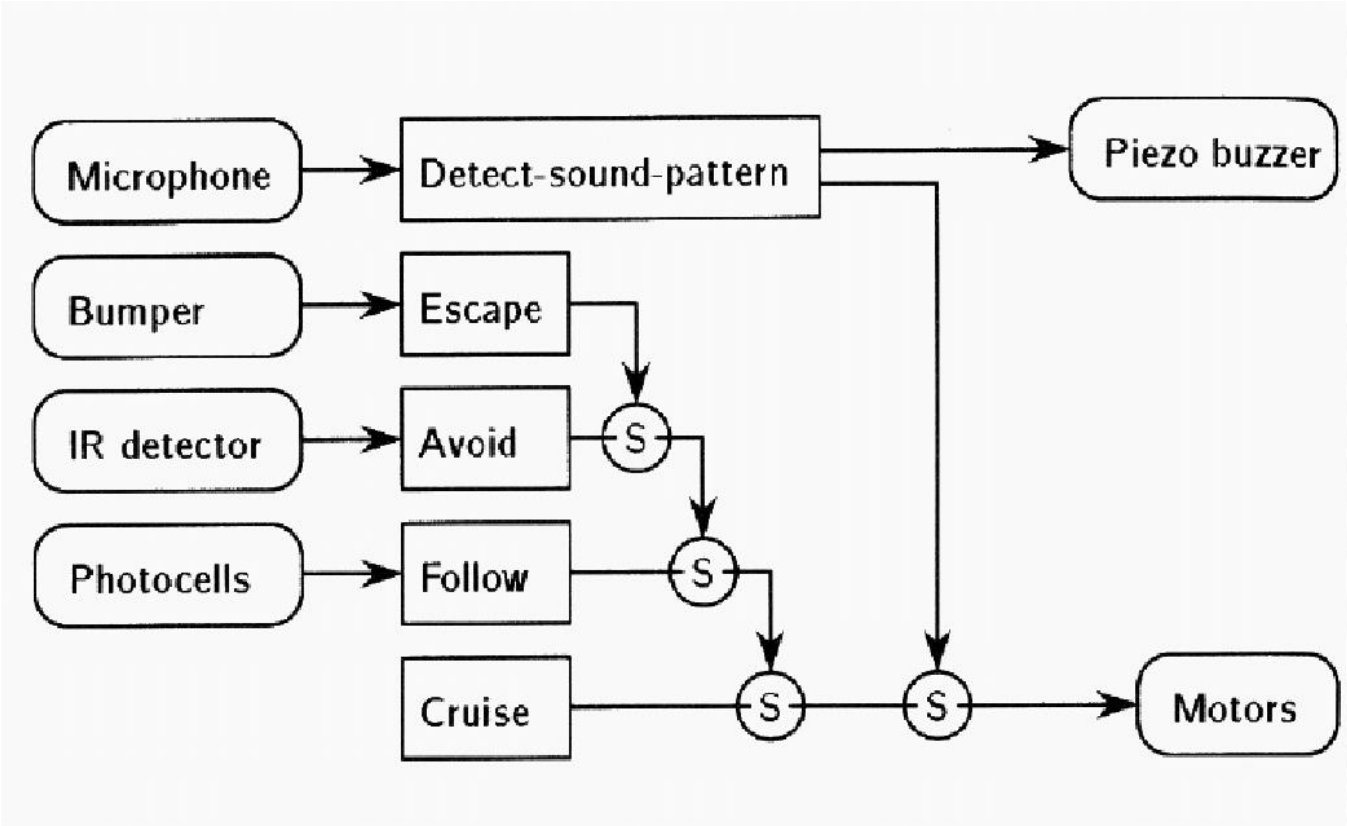
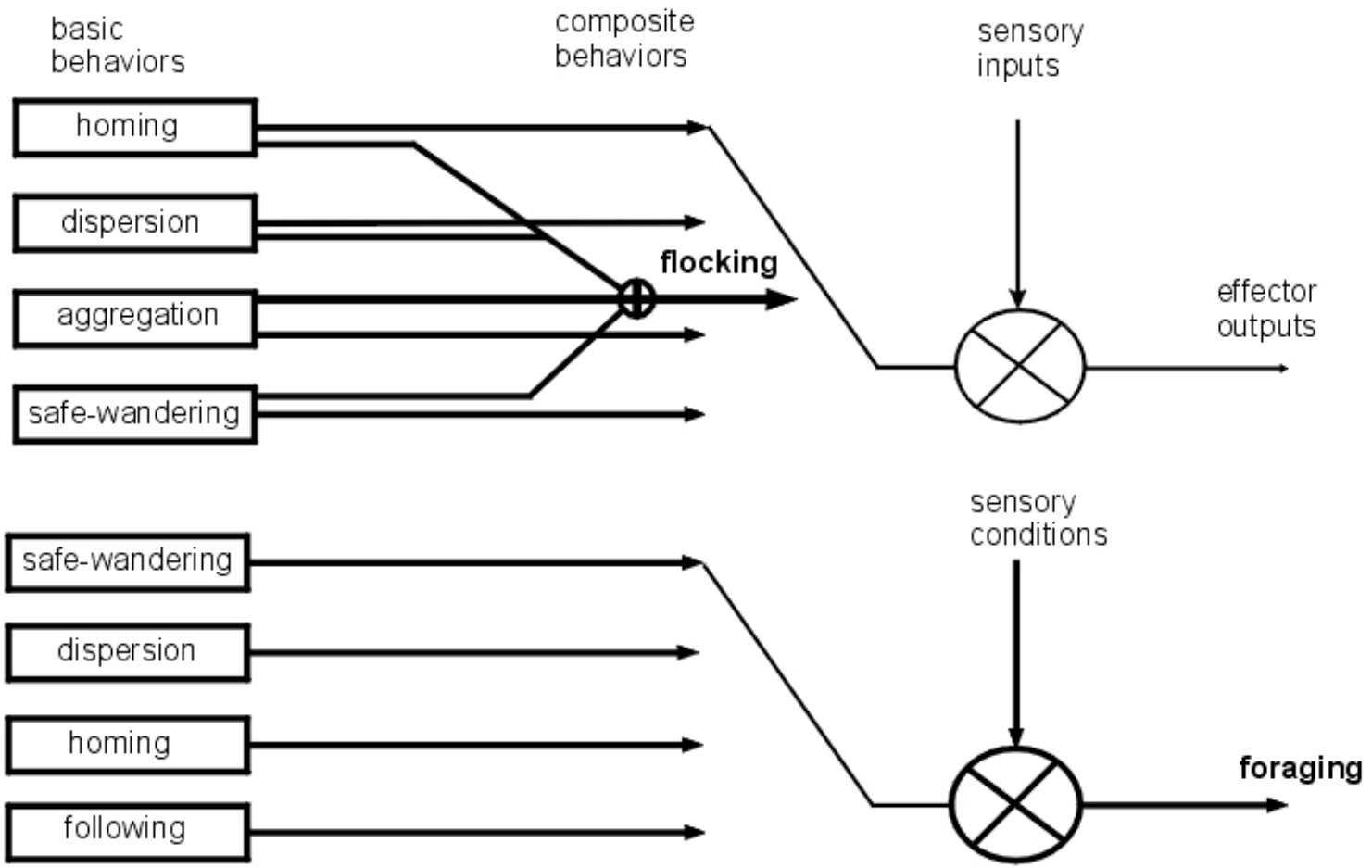
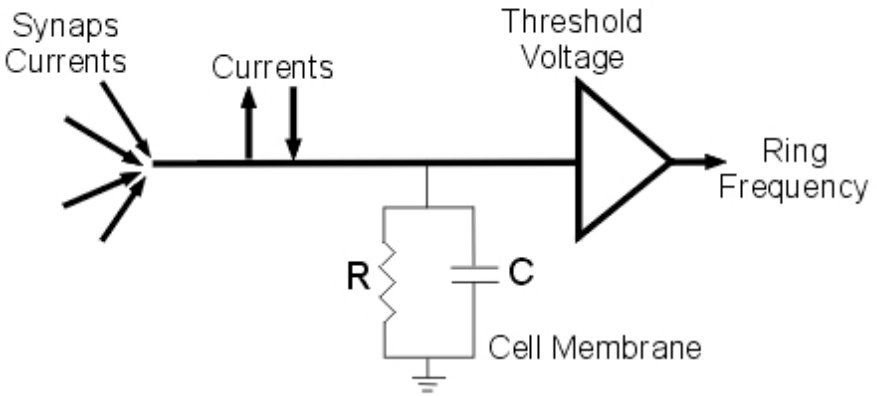
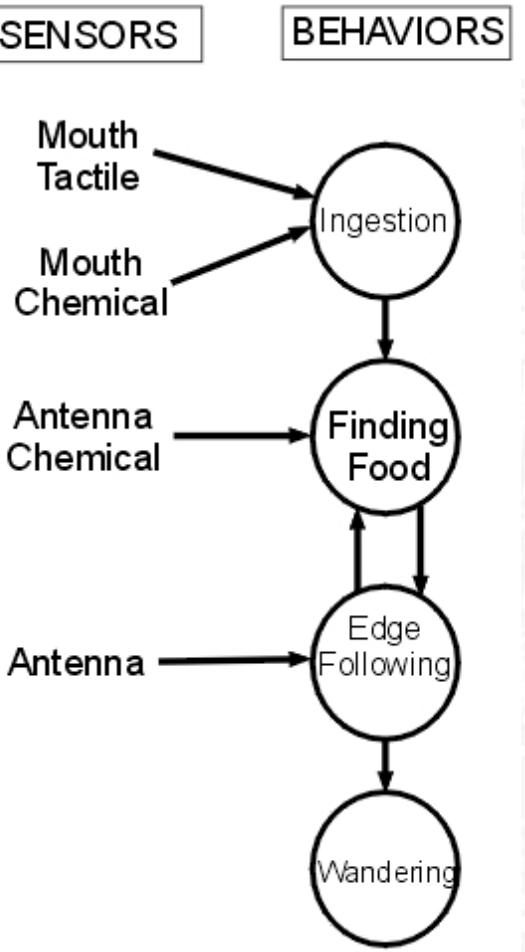


Foraging and Flocking



Cockroach Neuron / Behaviors





Steuerungsarchitektur eines Fisches

Control and information flow in artificial fish:

- Perception: sensors, focusser, filter
- Behaviours: behaviour routines
- Brain/mind: habits, intention generator
- Learning: optimization
- Motor: motor controllers, actuators/muscles

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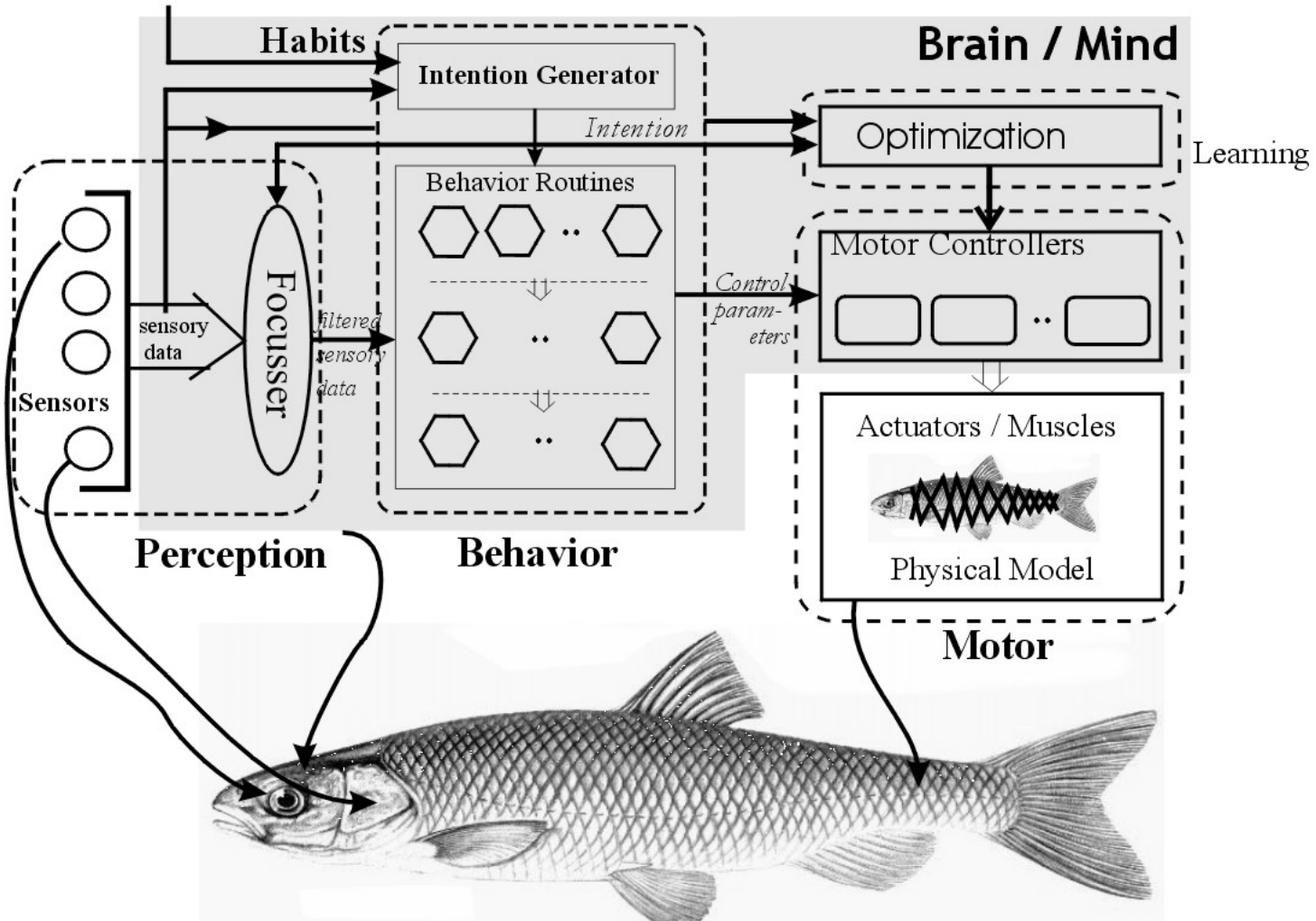
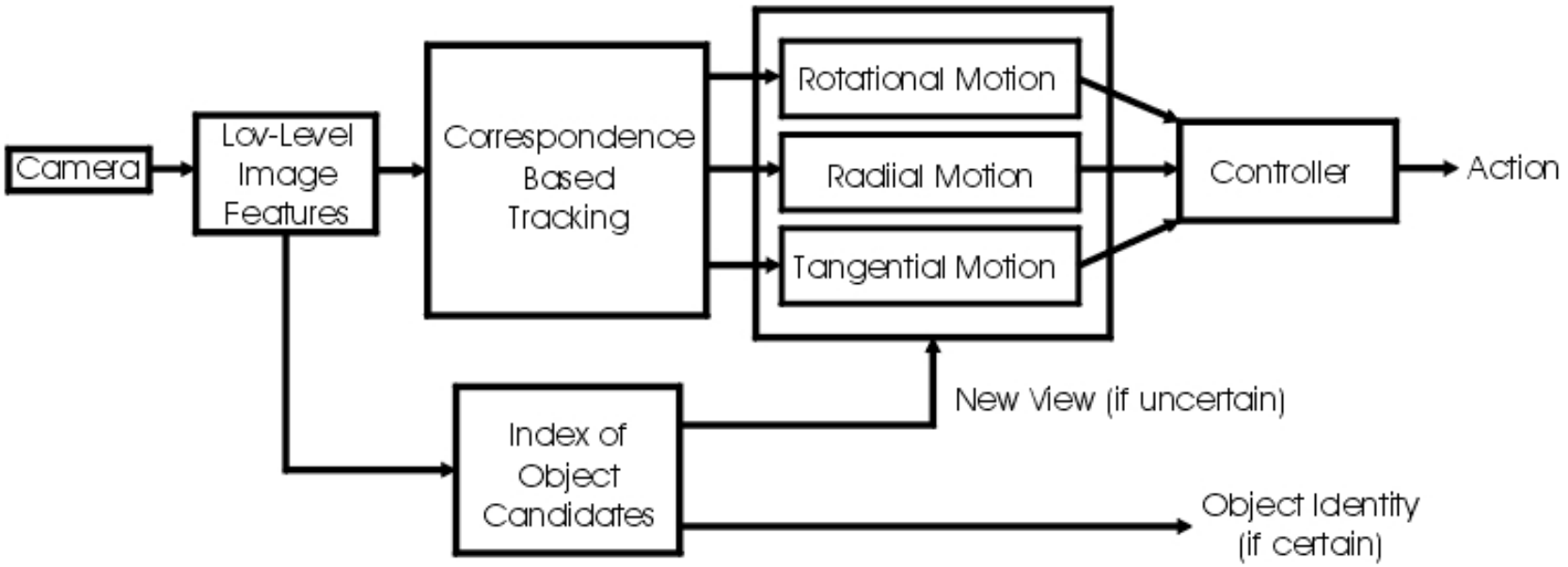
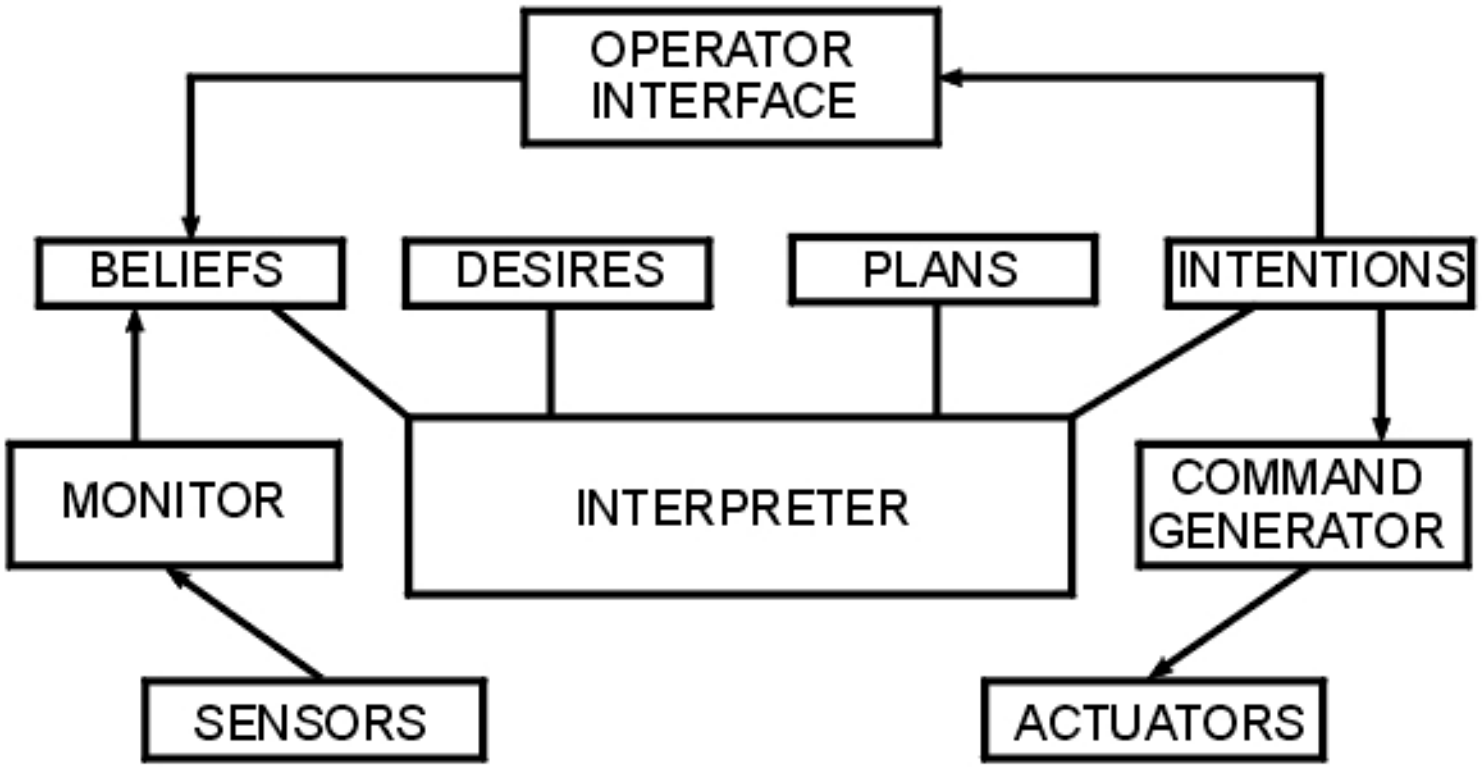


Active Vision Architecture

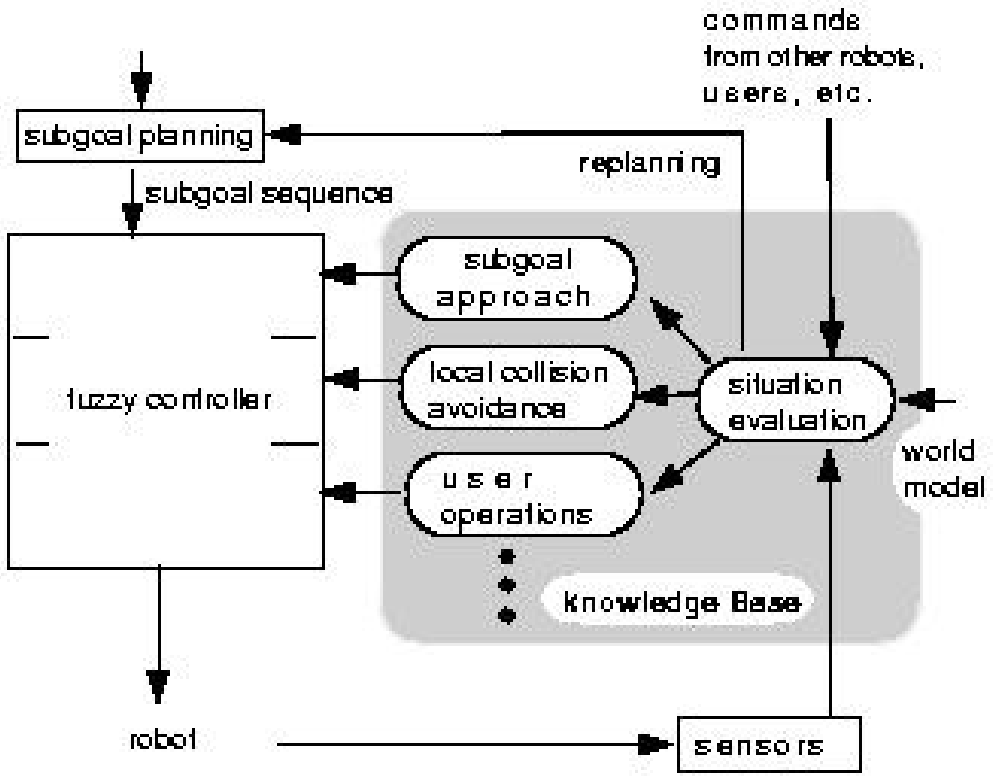


Procedural Reasoning System - SRI 1987



Verhaltensfusion

Hierarchische Fuzzy-Regelung eines Roboters



Verhaltensfusion (cont.)

Für die Situationsbewertung werden Fuzzy-Regeln benutzt

- Durch die Situationsbewertung werden 3 Fuzzy-Parameter bestimmt:
  1. die Priorität K
  2. der Replanning-Selector
  3. und *NextSubgoal*, ob ein Unterziel vorbei ist.
- Beispiel einer typischen Regel:

  IF (*SL85* IS HIGH) AND (*SL45* IS VL) AND (*SLR0* IS VL) AND (*SR45* IS VL) AND (*SR85* IS VL) THEN *K* IS HIGH AND *Replan* IS LOW
- Koordinierung von mehreren Regelbasen:

$$\begin{aligned} Speed &= Speed_{LCA} \cdot K + Speed_{SA} \cdot (1 - K) \\ Steer &= Steer_{LCA} \cdot K + Steer_{SA} \cdot (1 - K) \end{aligned}$$

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Hierarchie

A RCS (real-time Control System) reference model architecture for intelligent system. (Englisch, J. Albus)

- Processing modes are organised such that the BG (Behaviour Gernation) modules form a command tree.
- Information in the knowledge database is shared between SM (World Model) modules in nodes within the same subsystem. On the right, are examples of functional characteristics of SM modules and stored by the WM in the knowledge database at each level.

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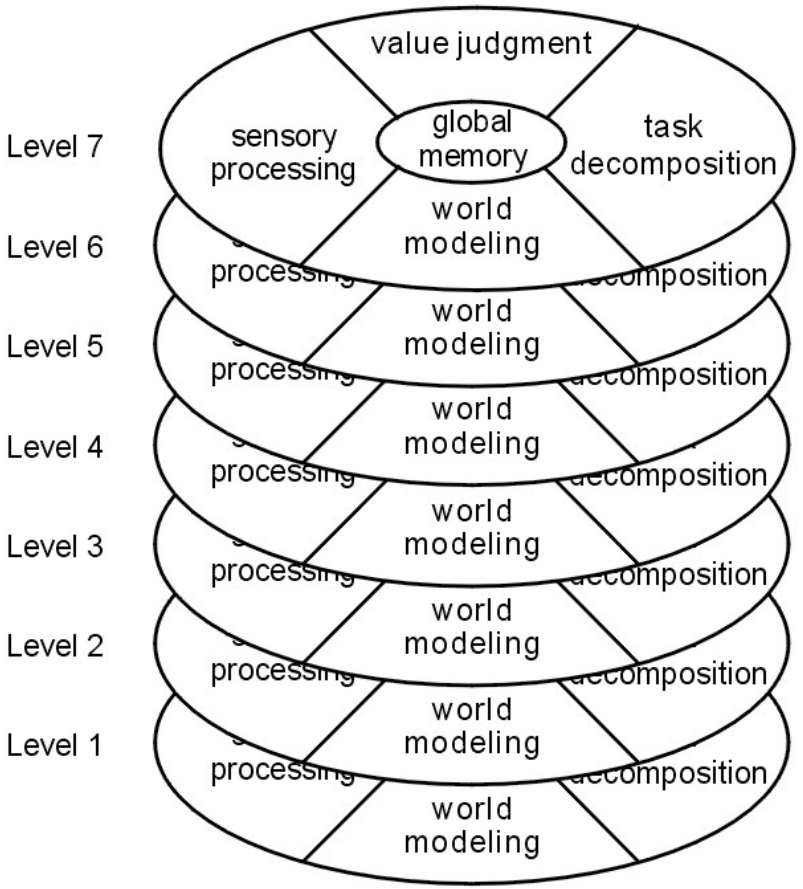

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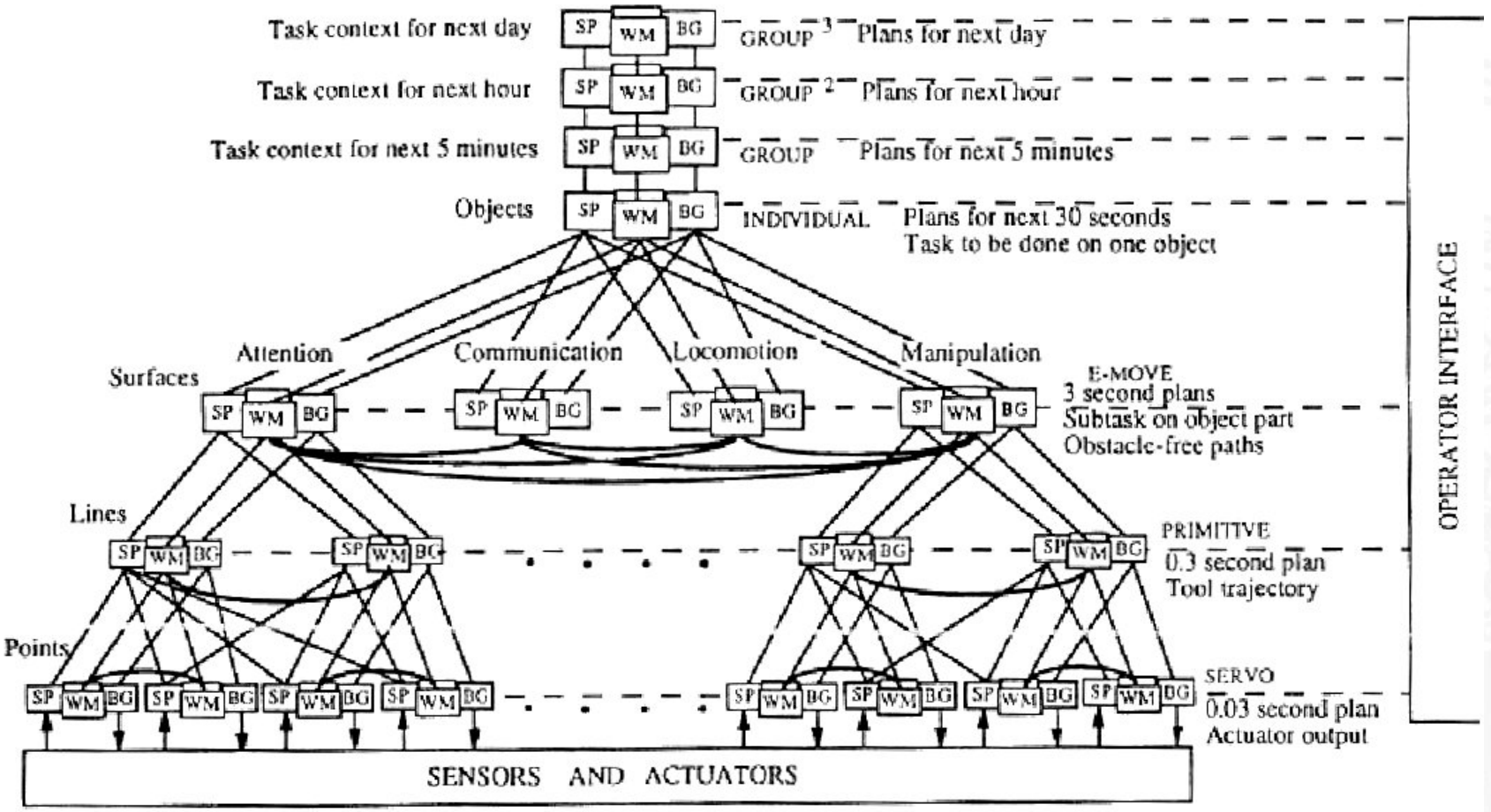

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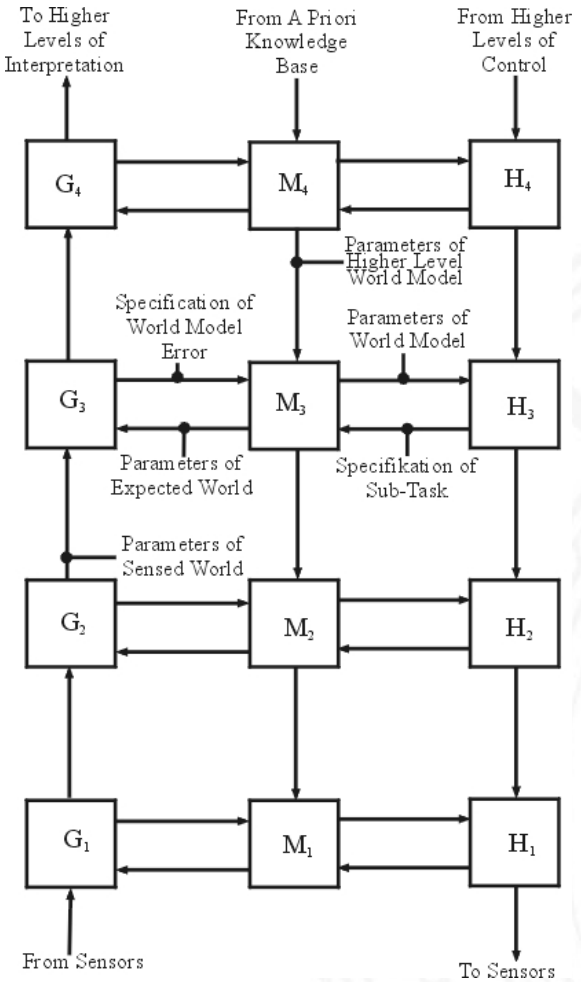
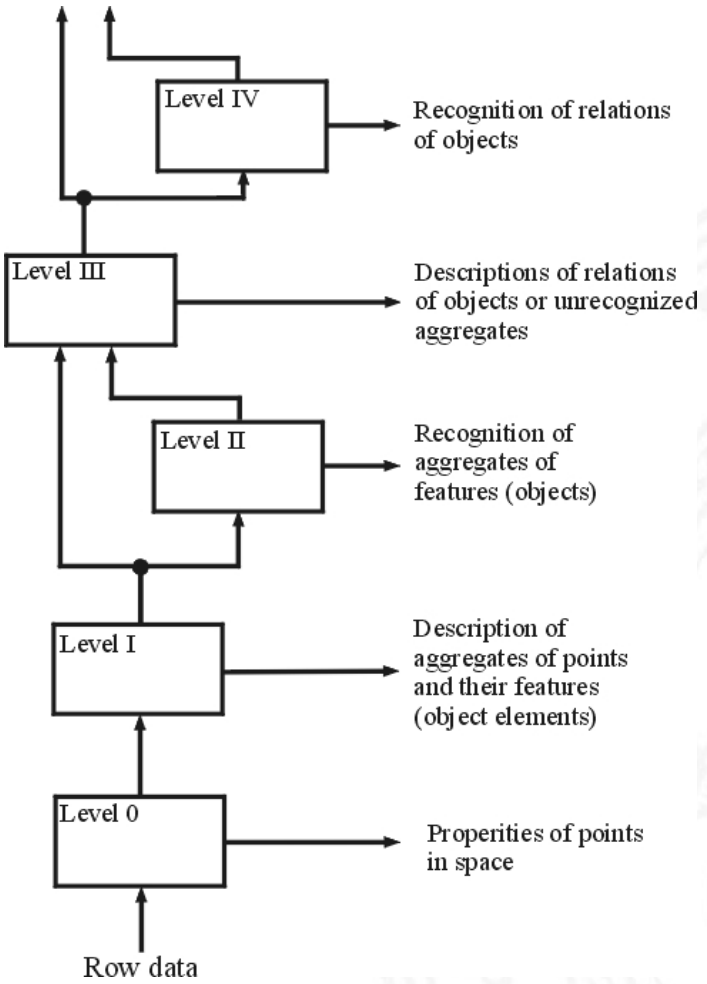
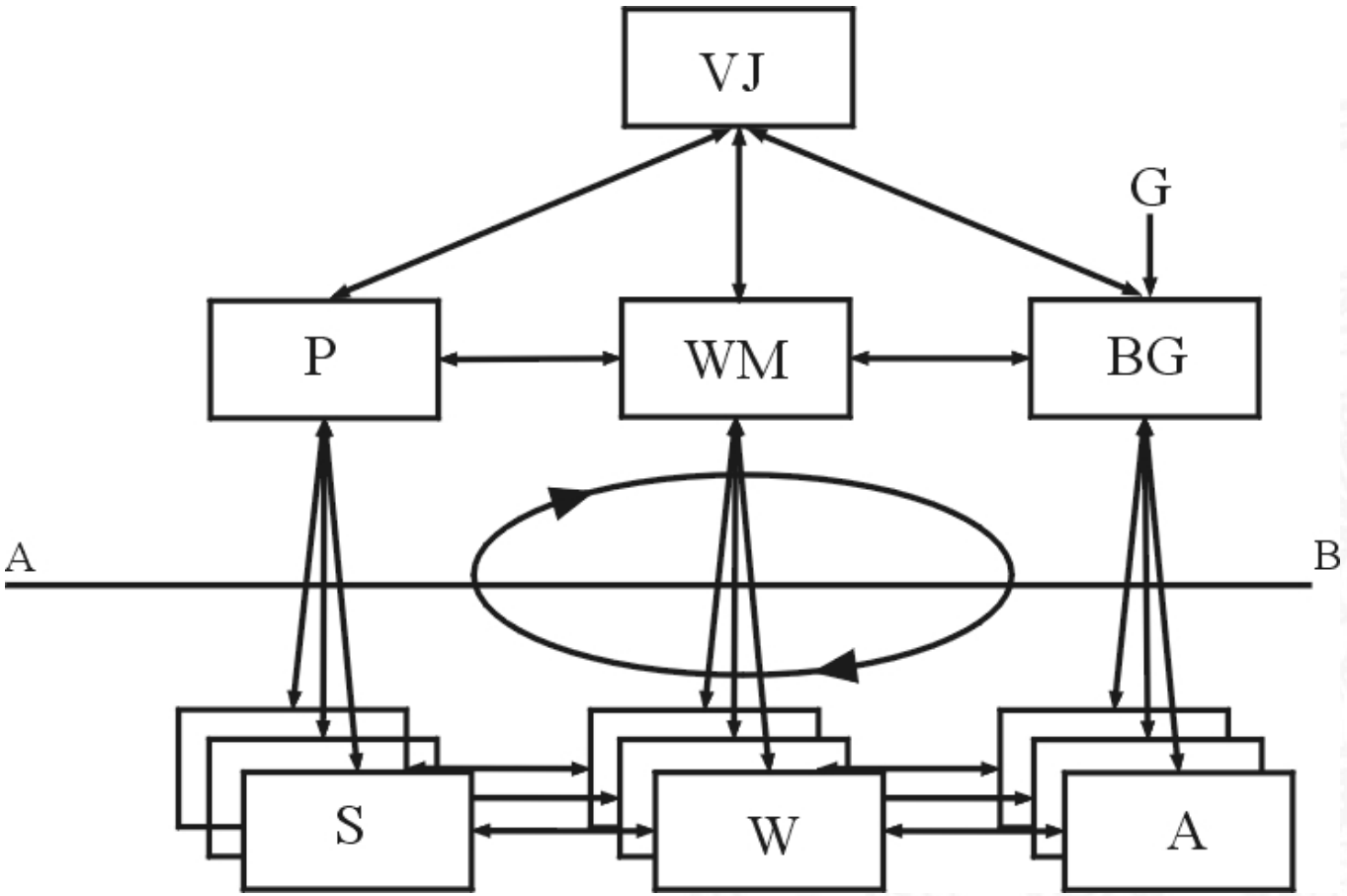


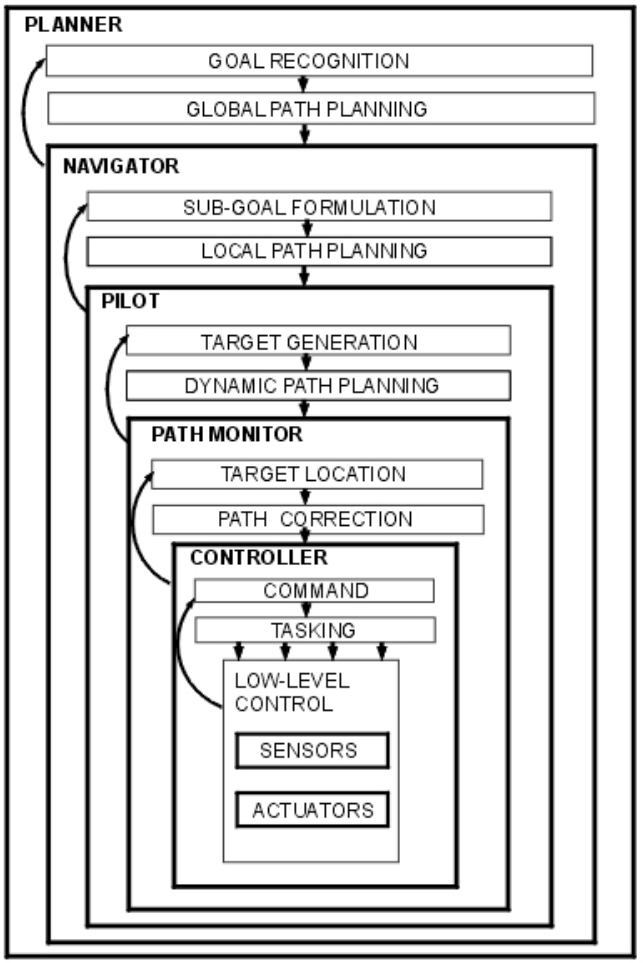
Sensor-Hierarchie



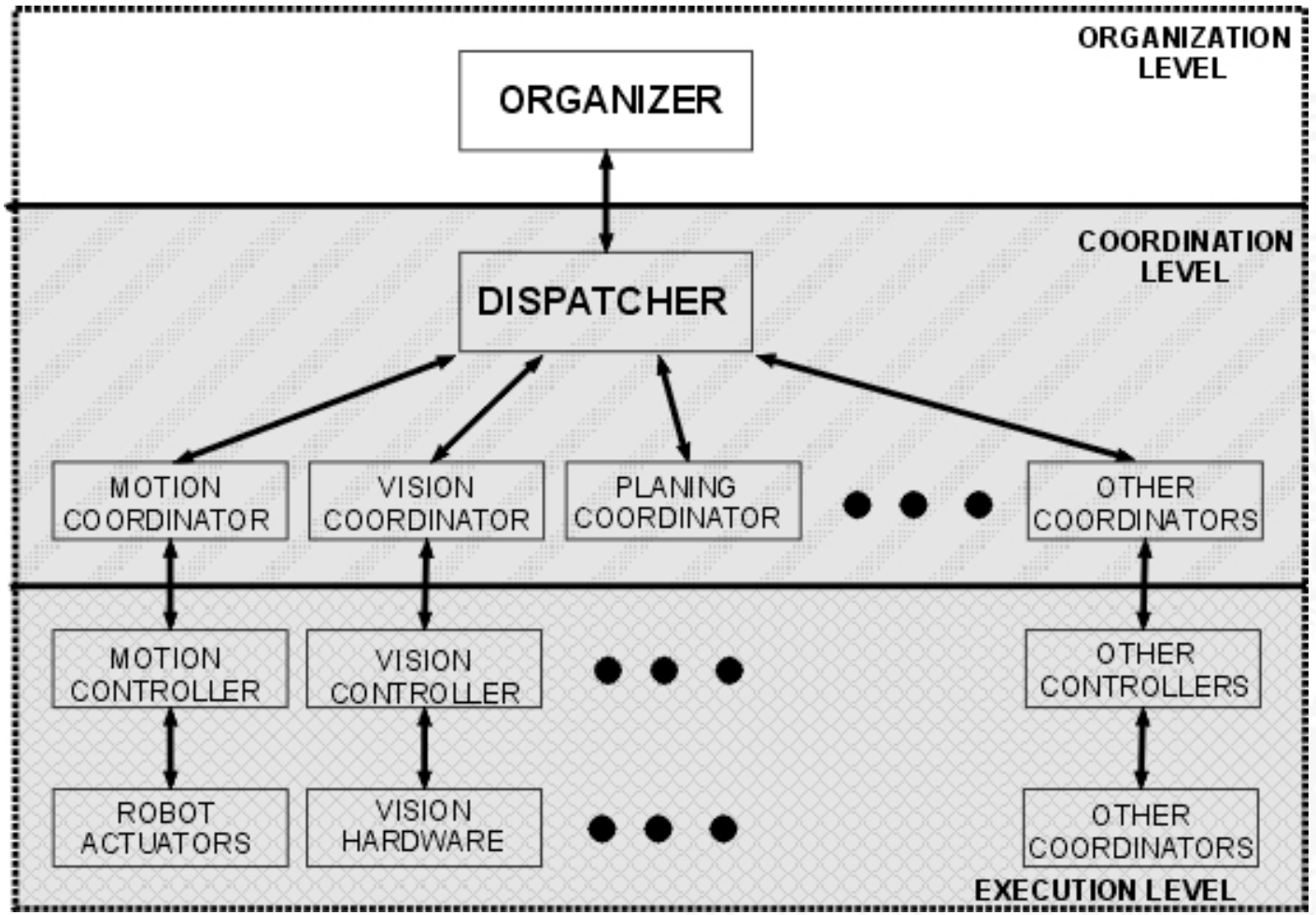
Hierarchie (cont.)



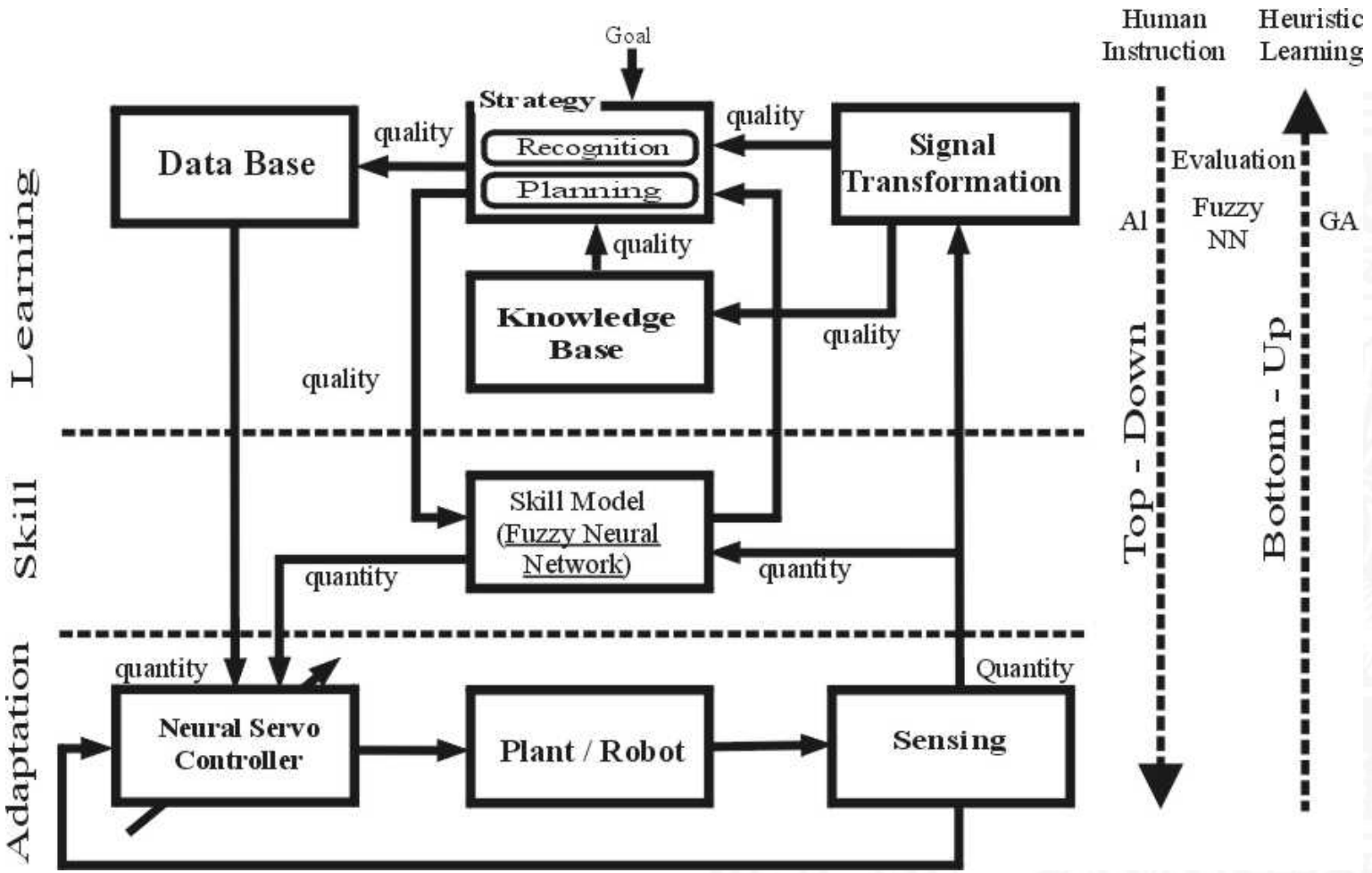
Ein anderes Beispiel: Meystel



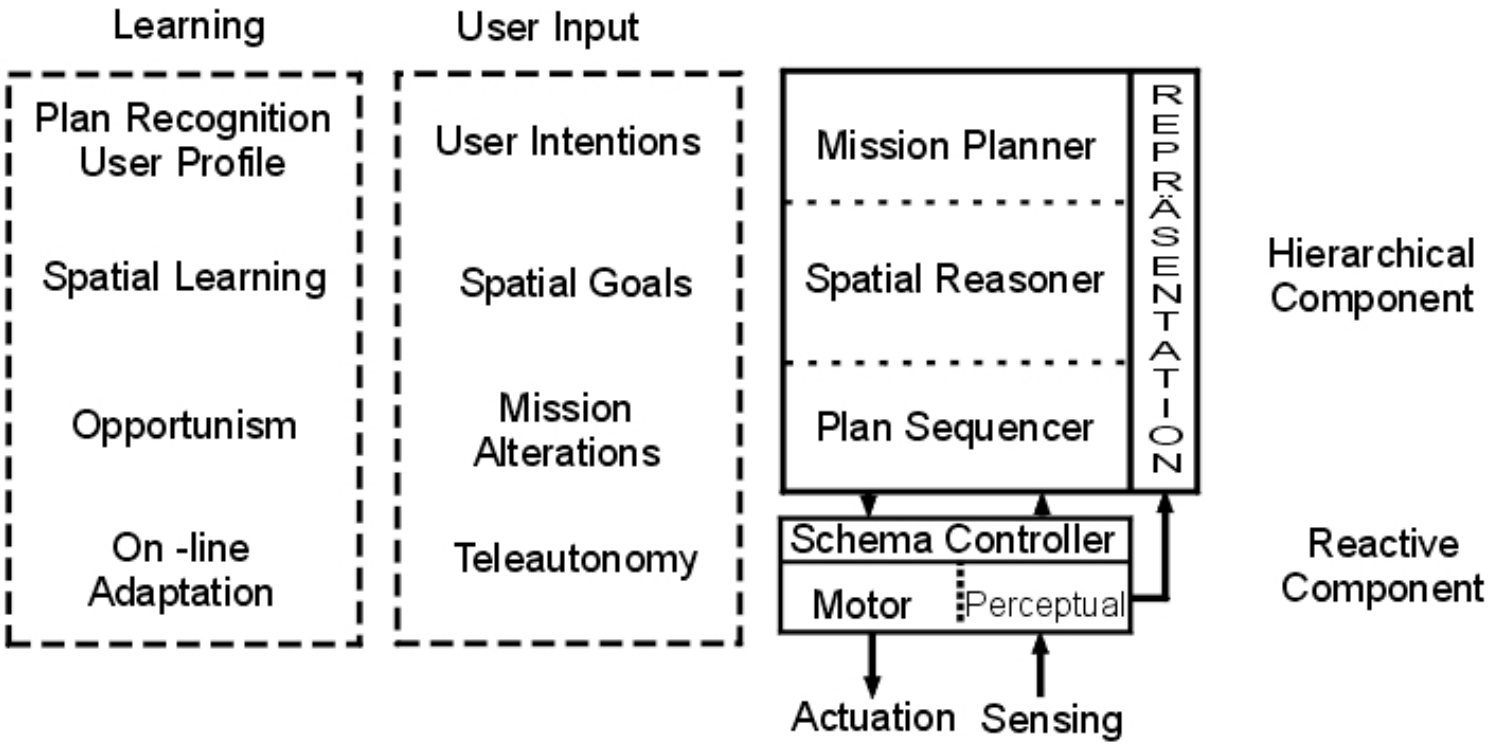
Ein anderes Beispiel: Saridis



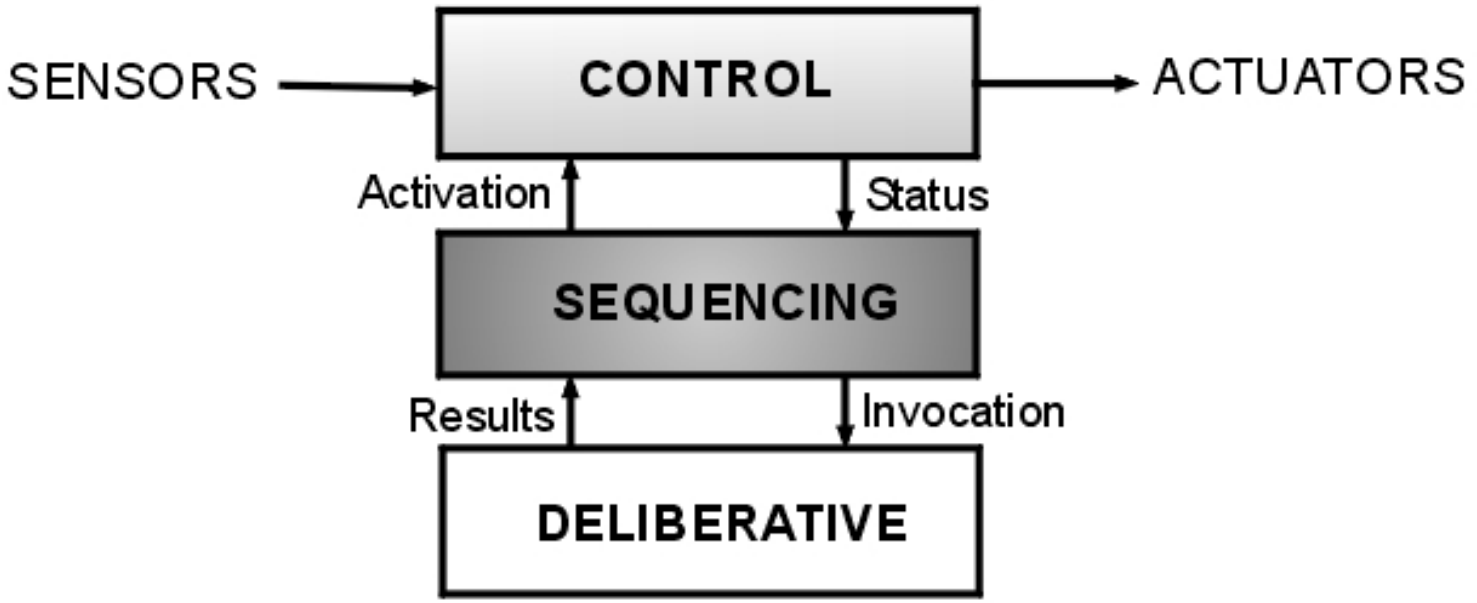
Eine Architektur für lernende Roboter



Das AuRA-Modell – Arkin '86

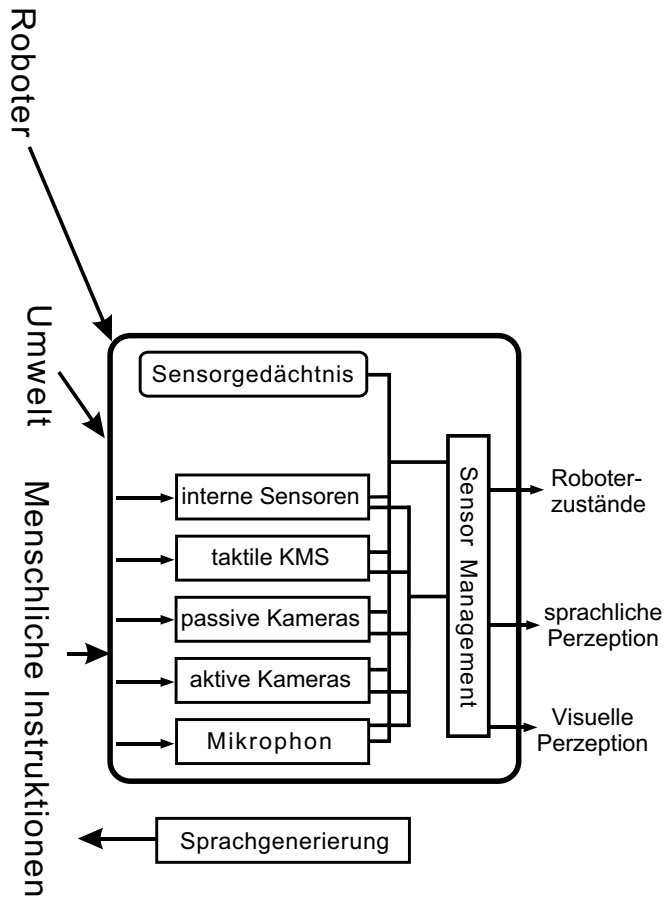


Atlantis – Nasa Rovers 1991



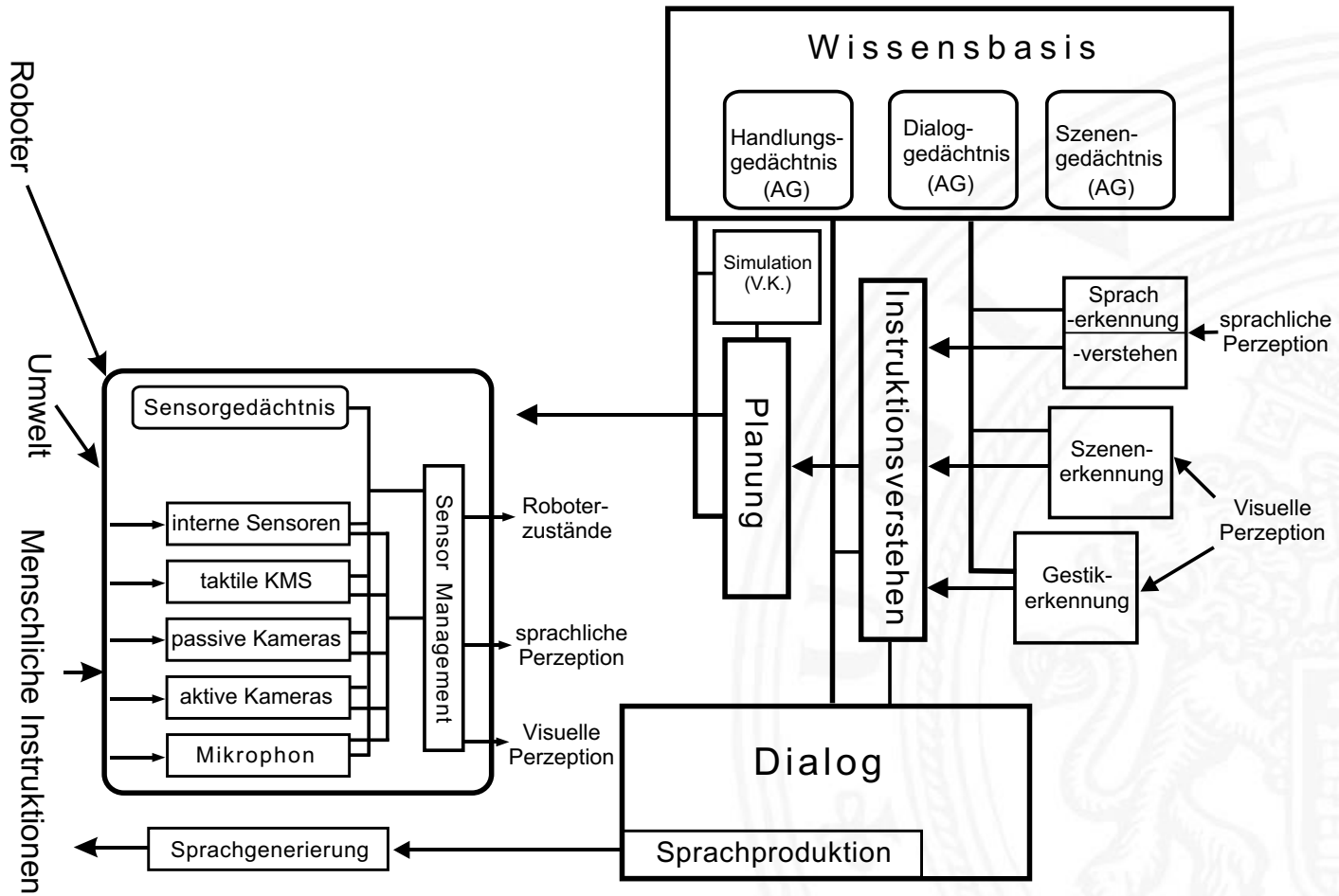
Architektur für interaktive Roboter

Perzeption



Architektur für interaktive Roboter (cont.)

+ Kognition



Architektur für interaktive Roboter (cont.)

Gesamtarchitektur