

# Scalable Distributed Intelligence



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# Motivation

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- As plants grow in **complexity** and control requirements grow in **difficulty**, there is an increased need for intelligence in control systems.
- **Intelligence**, traditionally a duty for central brains, has to be available where it is needed and that means, today, everywhere.
- This is achieved putting computing power and software in many places far from central brains. Intelligence is getting **ubiquitous** in control systems.
- Controllers are becoming societies.

# Summary

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- In this talk I will address the implementation of **complex controllers with distributed intelligence**
- Describe some **basic technologies** and some of the **trends**
- Propose **research topics** for future pervasive, highly intelligent, scalable controllers

# Contents

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## □ A **Fast Review**

- Distributed Control
- Distributed Artificial Intelligence
- Distributed Intelligent Control

## □ Some **Trends and Speculations**

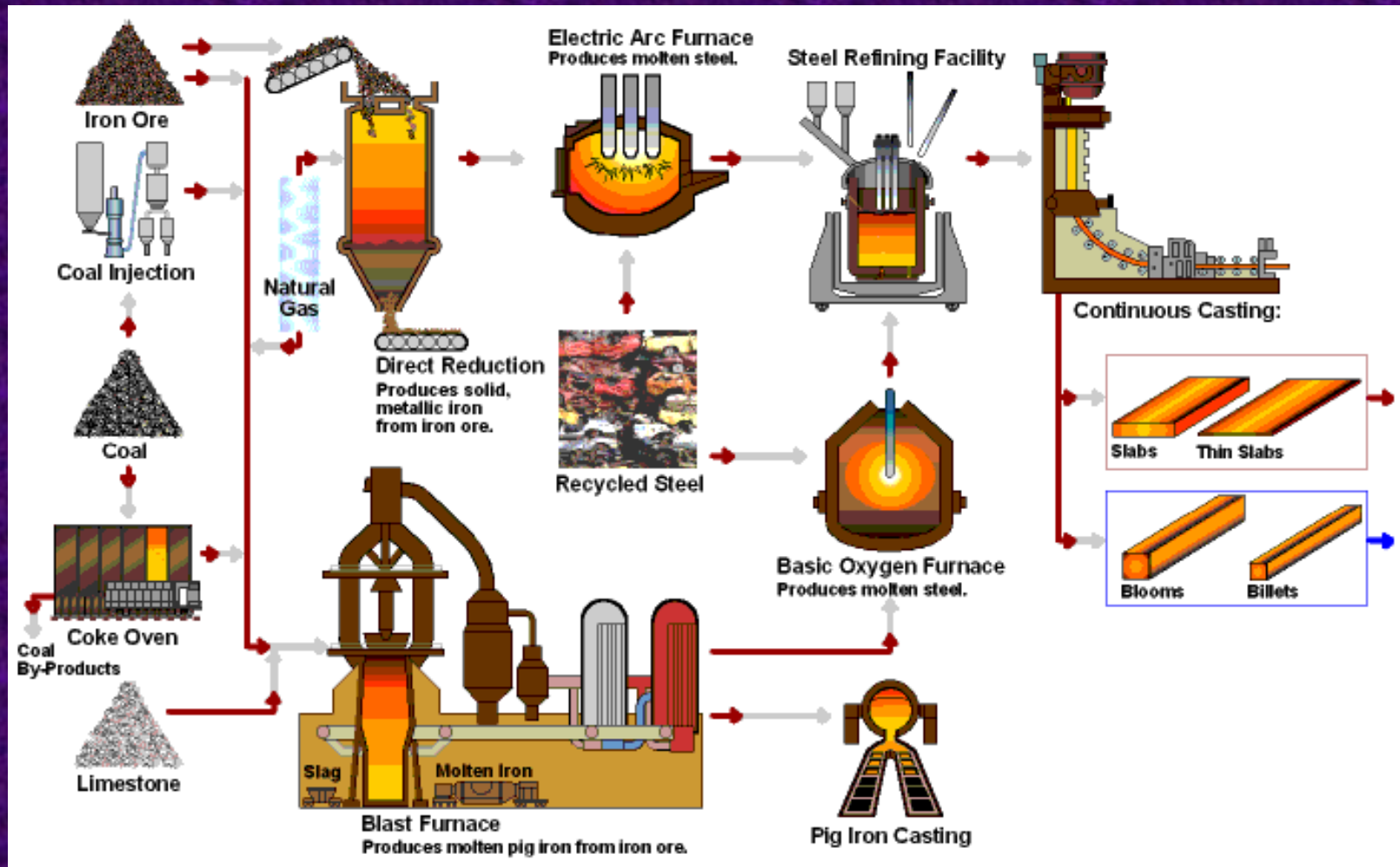
- Deeper Intelligence
- Scalable Intelligence
- Scalable Distributed Intelligence

# Distributed Control



When the control pieces are  
taken apart

# Distributed Plants

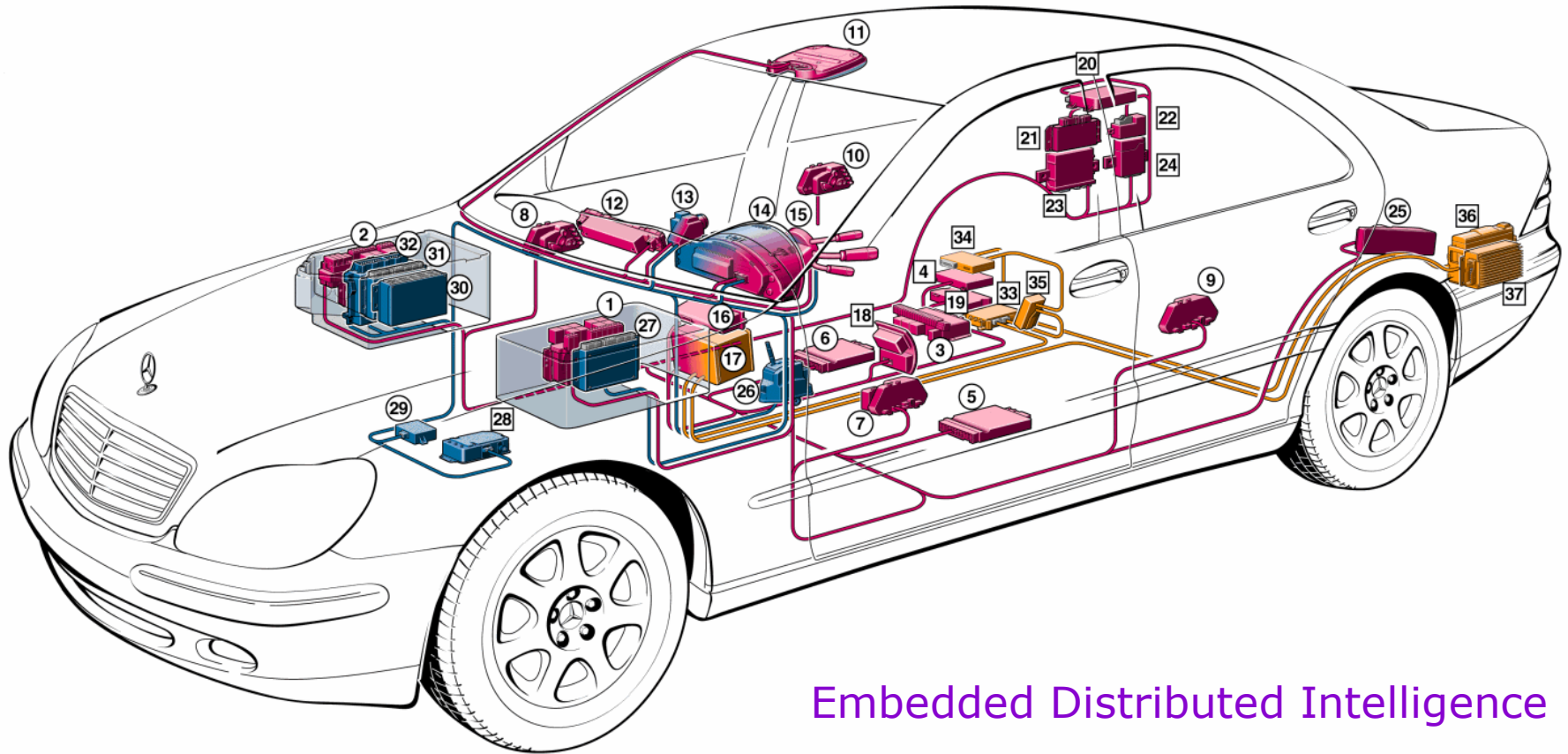


# Even in small size

CAN Class B

CAN Class C

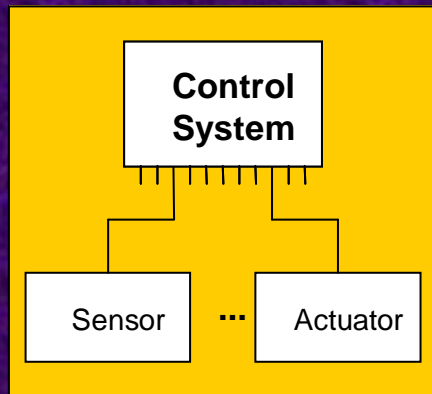
D2B optical



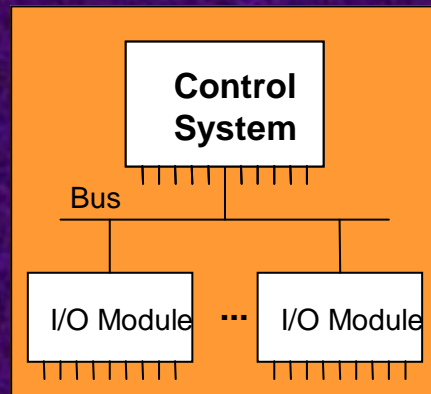
Embedded Distributed Intelligence

# Trends

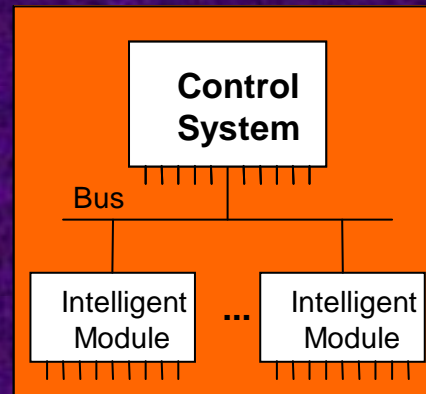
Trend towards distributed systems



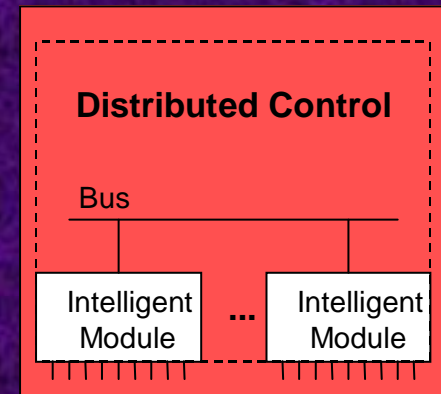
Central I/O



Decentralized I/O



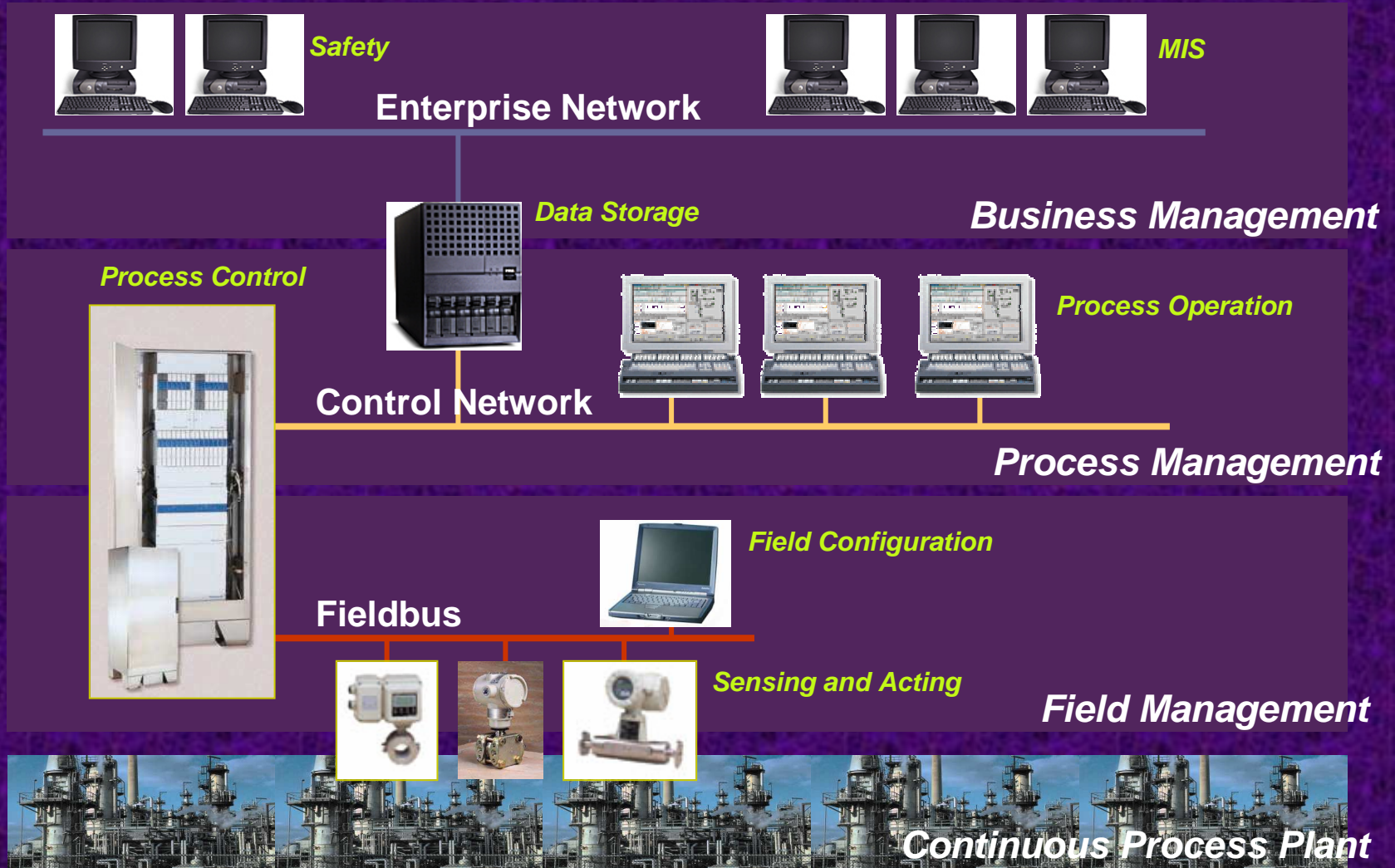
Intelligent Devices



Distributed Systems

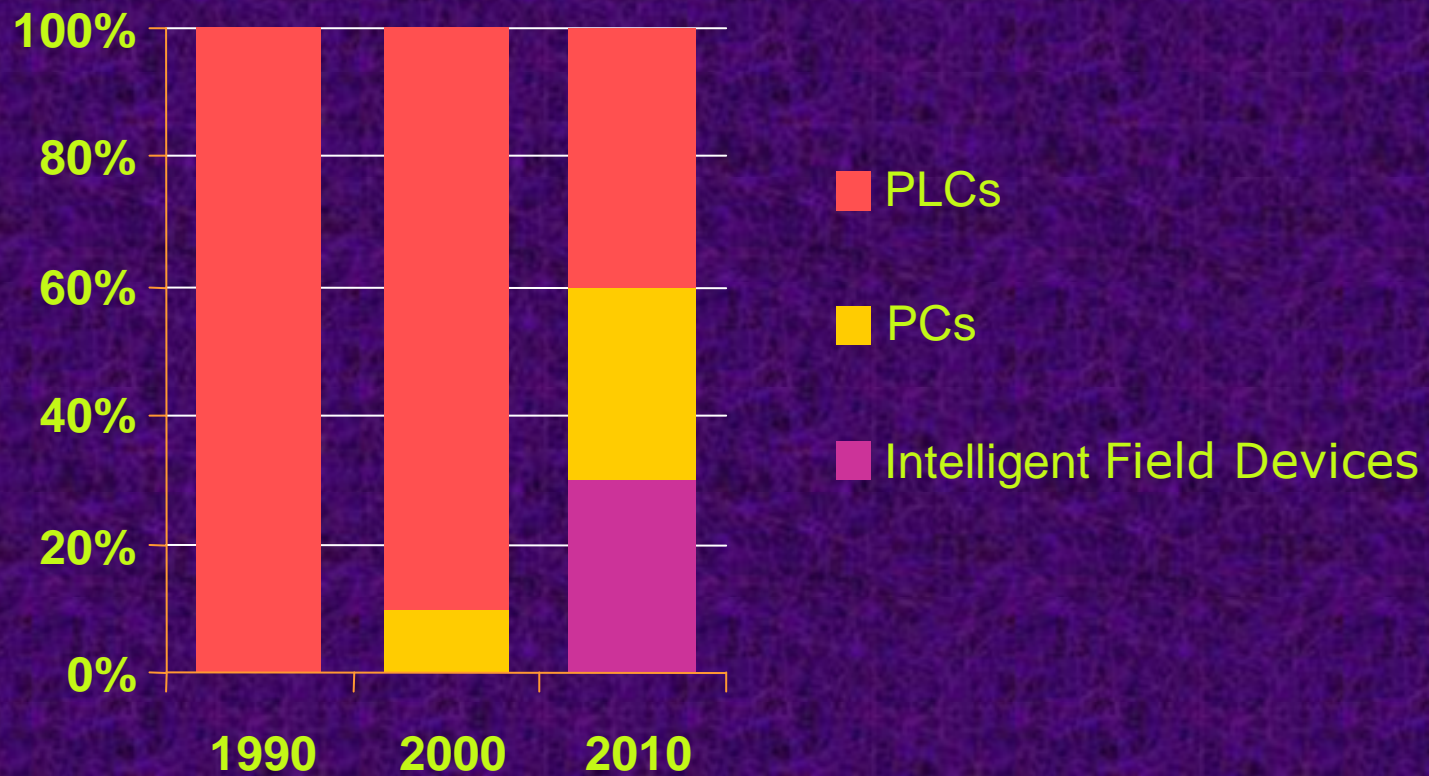


# Plant-wide Control



# Perspectives

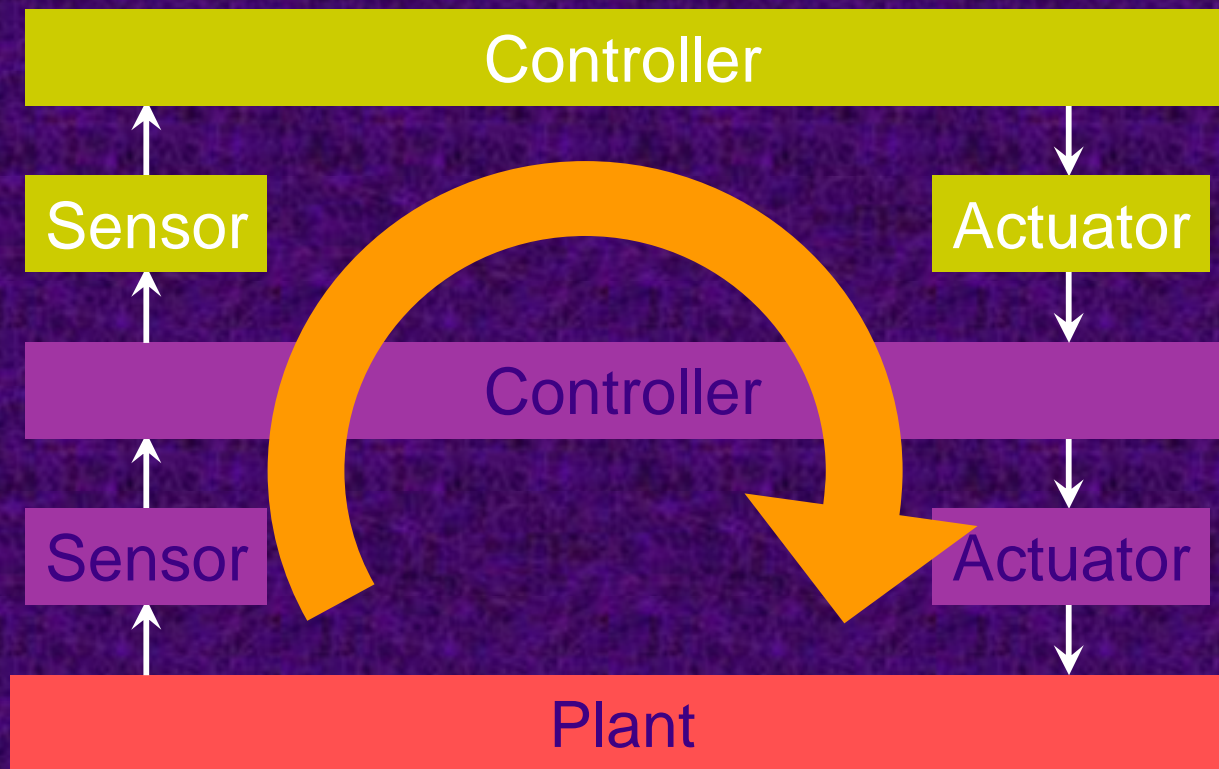
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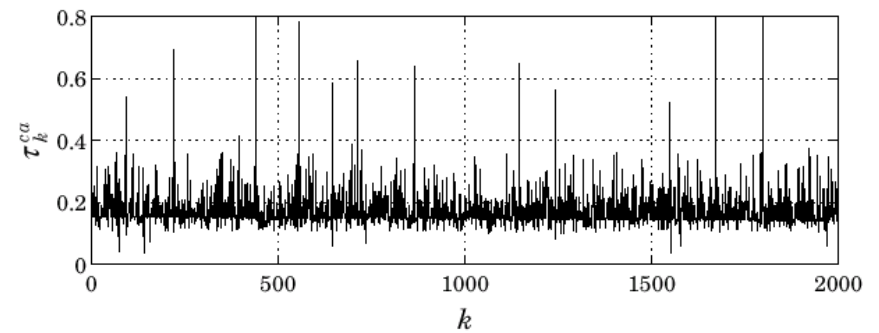
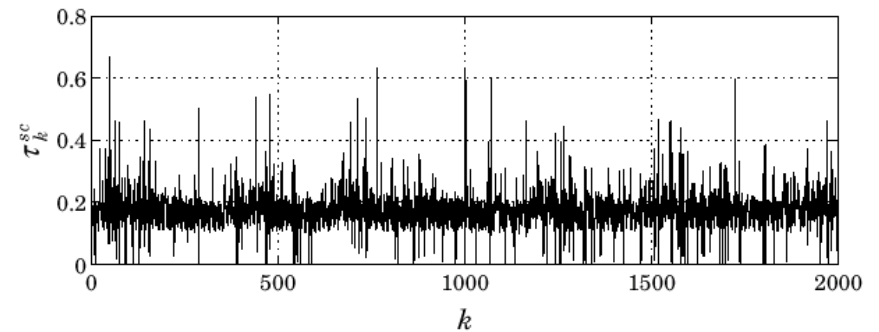
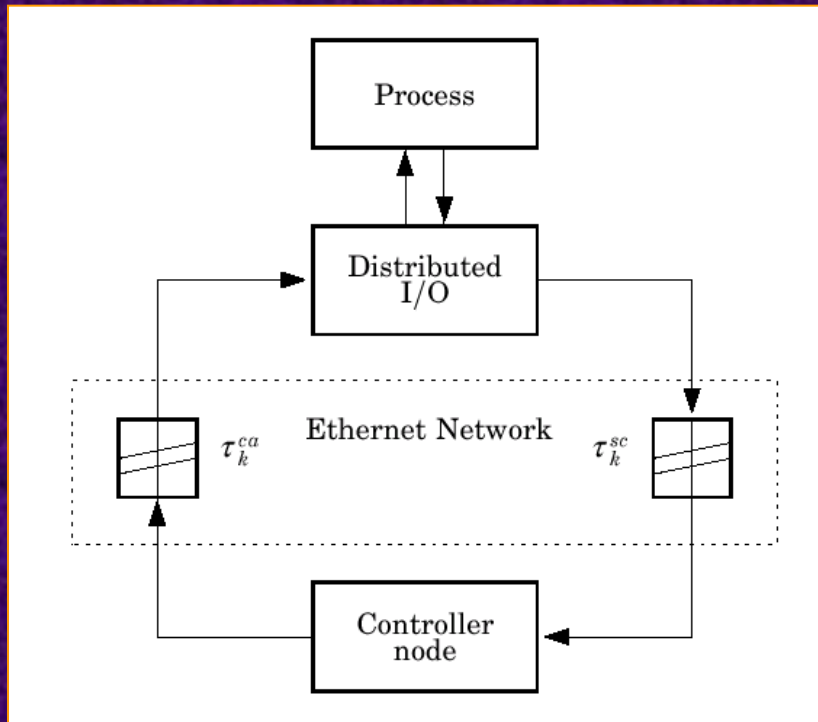
[PNO 2001]

# Closing loops through the net

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# Network Control



# Design Challenges

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- Predictability
- Performance
- Scalability
- Openness
- Coordination
- Transparency
- Naming
- Cost
- Consistency
- Failure handling
- Security
- Heterogeneity
- Mobility
- Load sharing
- Footprint
- *etc.*

# Where's the infrastructure ?

PROGRAM	TOPIC	STD CANDIDATES
<b>ANTS</b>	Negotiation-based resource management	Architectural framework Negotiation protocols
<b>NEST</b>	Robust, compact coordination services for large-scale distributed embedded systems	Coordination service API-s Composable, lean micro-middleware
<b>PCES</b>	Language and compiler technology for DRE-s	Aspect/weaving API-s Componentized/aspectified middleware
<b>ARMS</b>	Adaptive,reflective middleware for simultaneous control of multiple QoS	QoS API-s ORB services to support QoS
<b>SEC</b>	High-performance control algorithms and systems	Controls API (facility) Design patterns for assurance
<b>MoBIES</b>	Model-based Integration of Embedded Systems via domain-specific tools	Domain specific modeling languages
<b>NEPHEST</b>	National Experimental Platform for Hybrid and Embedded Systems	A relatively new, long-term NSF program

[Watson]

# Distributed Artificial Intelligence



When the network thinks

# What's intelligence?

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- The capability of exploit knowledge
- What's knowledge then?
  - Machine and task specific meaningful information



# Why distribute intelligence?

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- Intelligence is distributed across the plants for many reasons (both needs and enablers):
  - availability of suitable embeddable processors,
  - timing requirements that forbid communication due to latencies,
  - need of increased levels of performance that is achieved through parallelism,
  - simplification of construction and maintenance tasks through modularity,
  - reduction of cost and time-to-market by means of component-based reuse,
  - integration of legacy systems
  - availability of software platforms for AI
  - ...

# Distributed Artificial Intelligence

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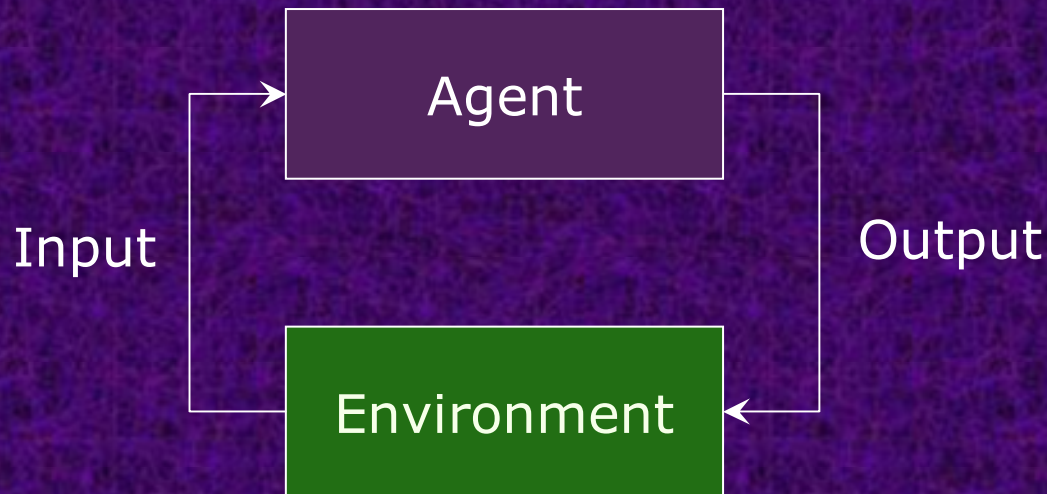
- **Distributed Problem Solving (DPS)**
  - Subalgorithms that collaborate
  - E.g. blackboard systems
  
- **Parallel AI (PAI)**
  - Splitting an algorithm for performance
  - Typically on multiprocessors
  
- **Multi-agent systems (MAS)**
  - When each component thinks

[Nwana]

# A Definition of Agent

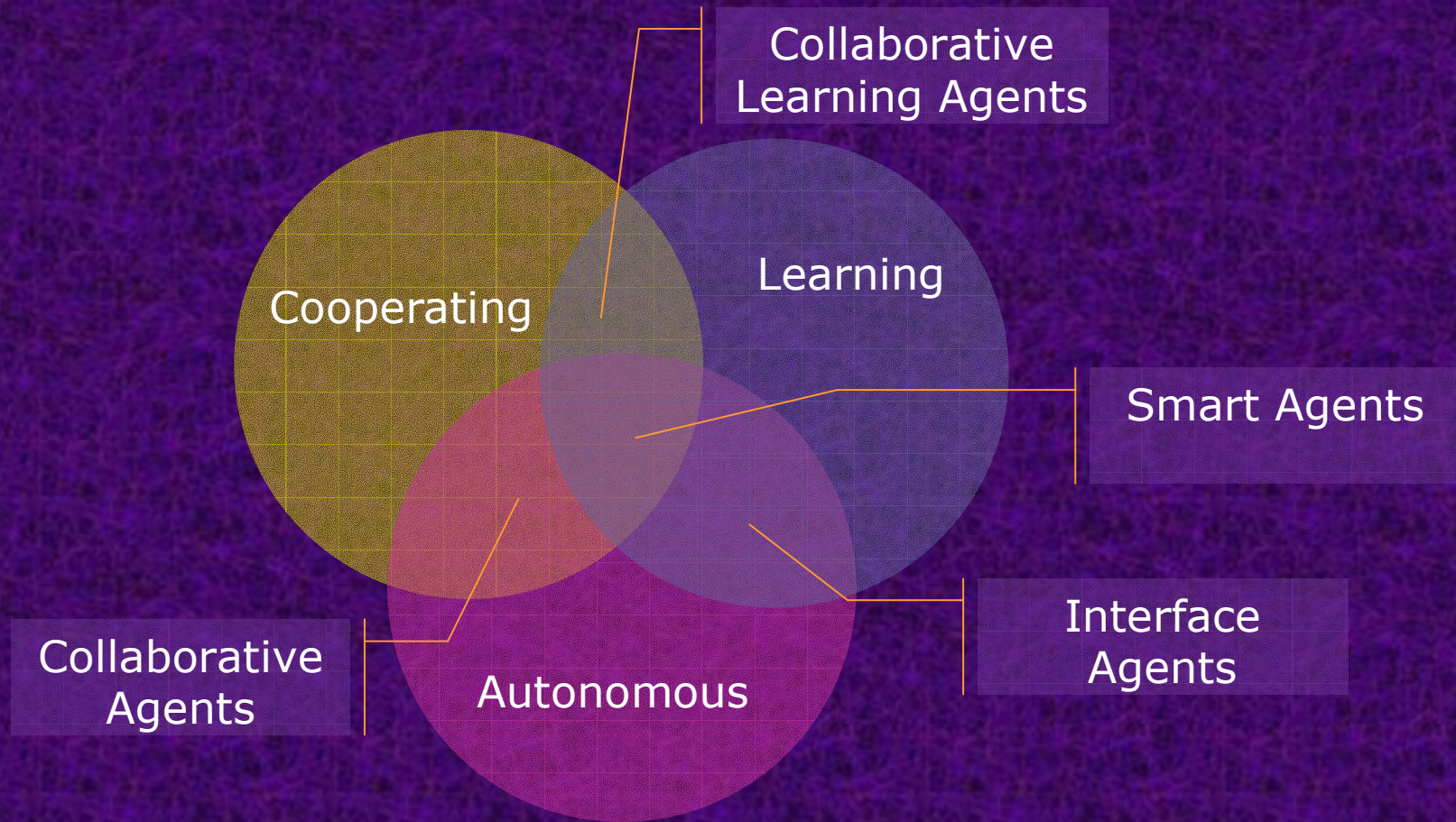
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- An agent is an autonomous computer-based system capable of independent action in some environment with control over its internal state.



# Agent Typology

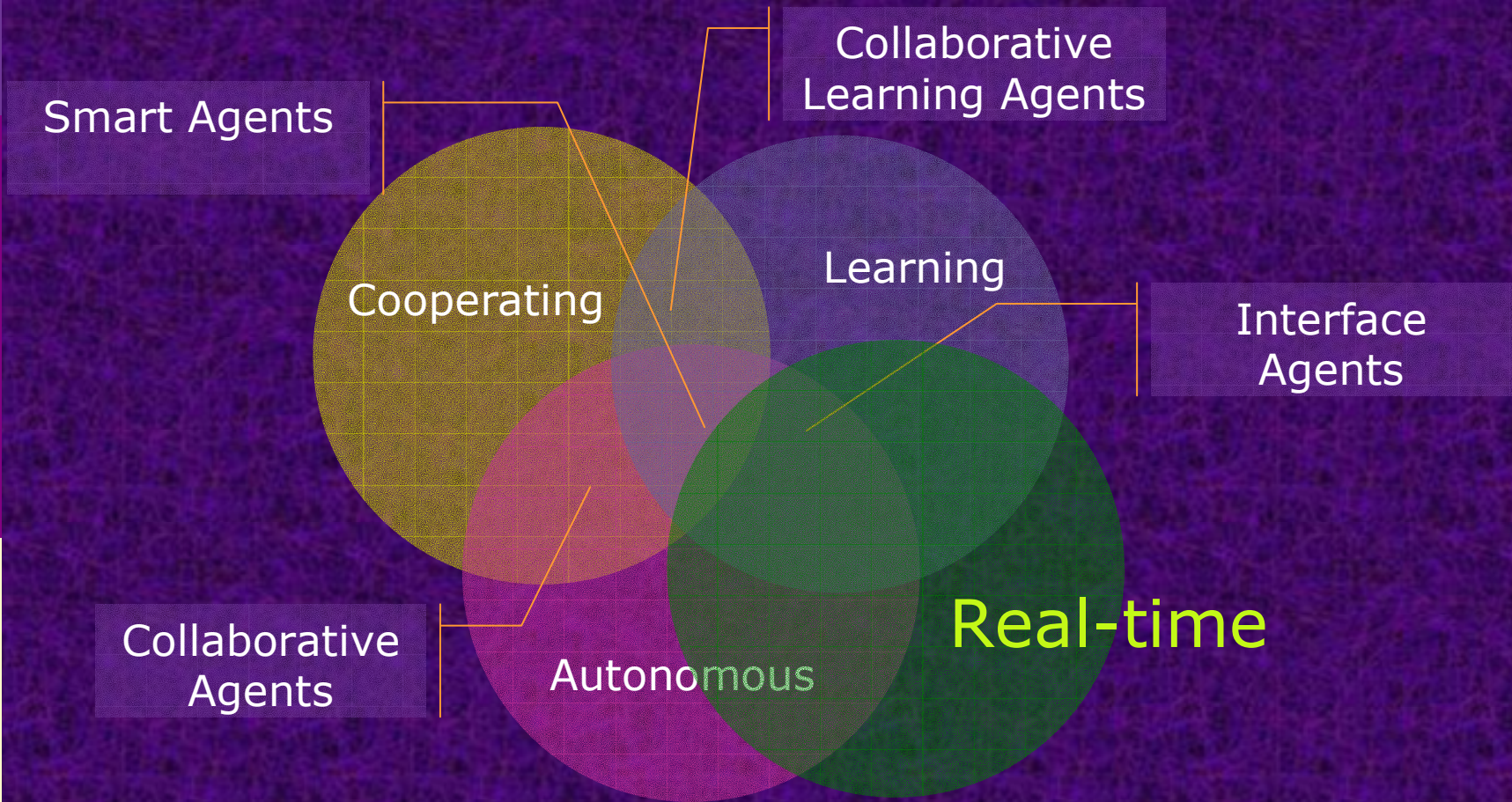
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[Nwana]

# Agent Typology

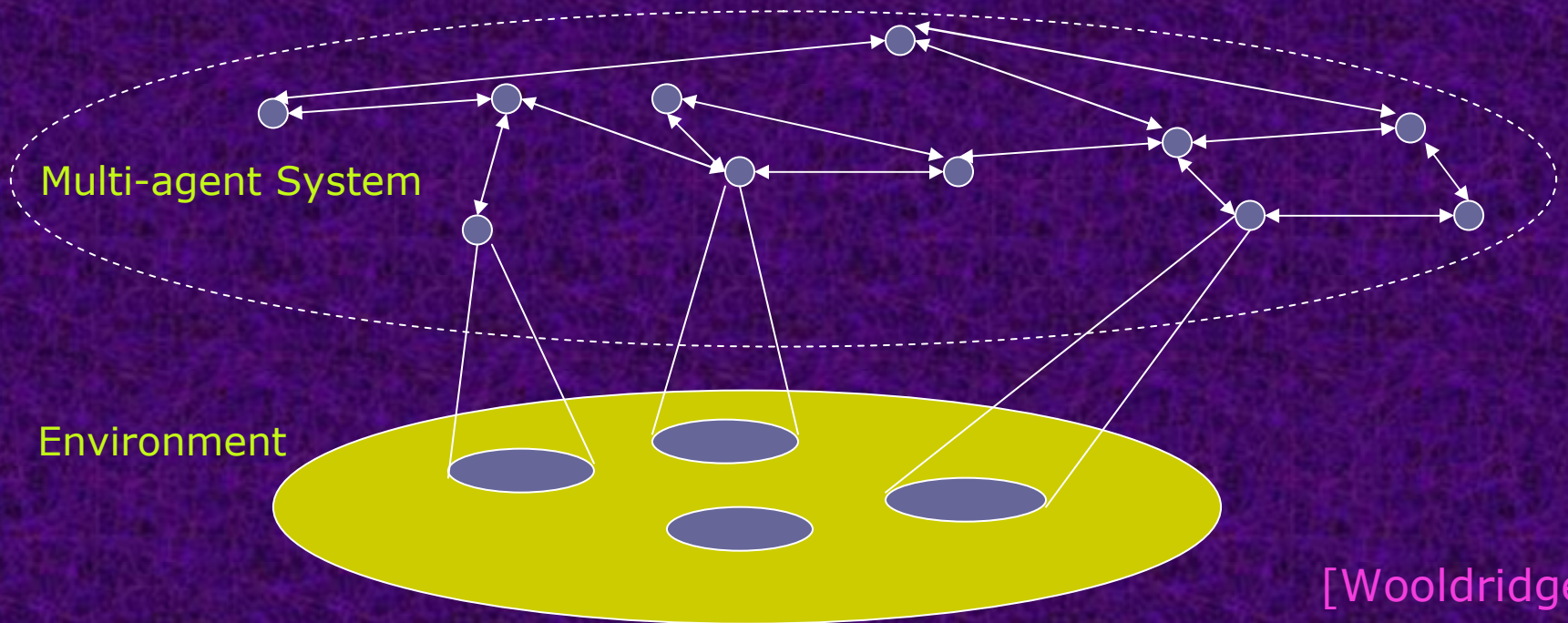
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[Nwana]

# Multi-agent Systems (MAS)

A number of agents which interact with one another through communication. Each agent will act upon or influence different parts of the environment.



# Distributed Intelligent Control

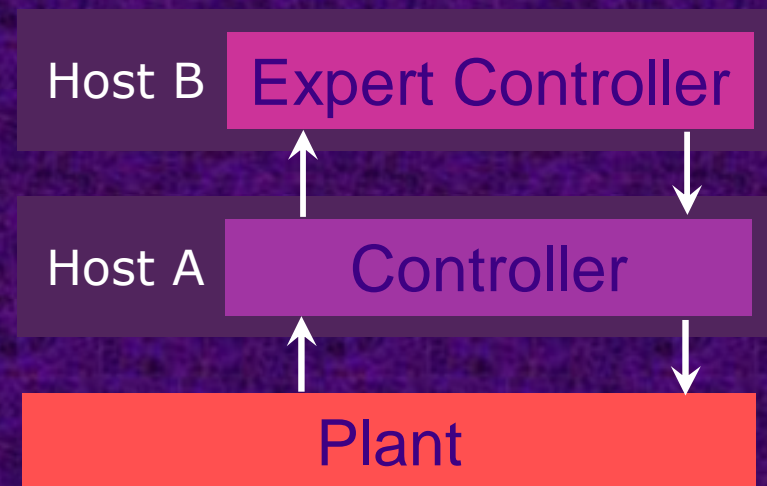


Taking apart control thought  
processes

# Elementary case

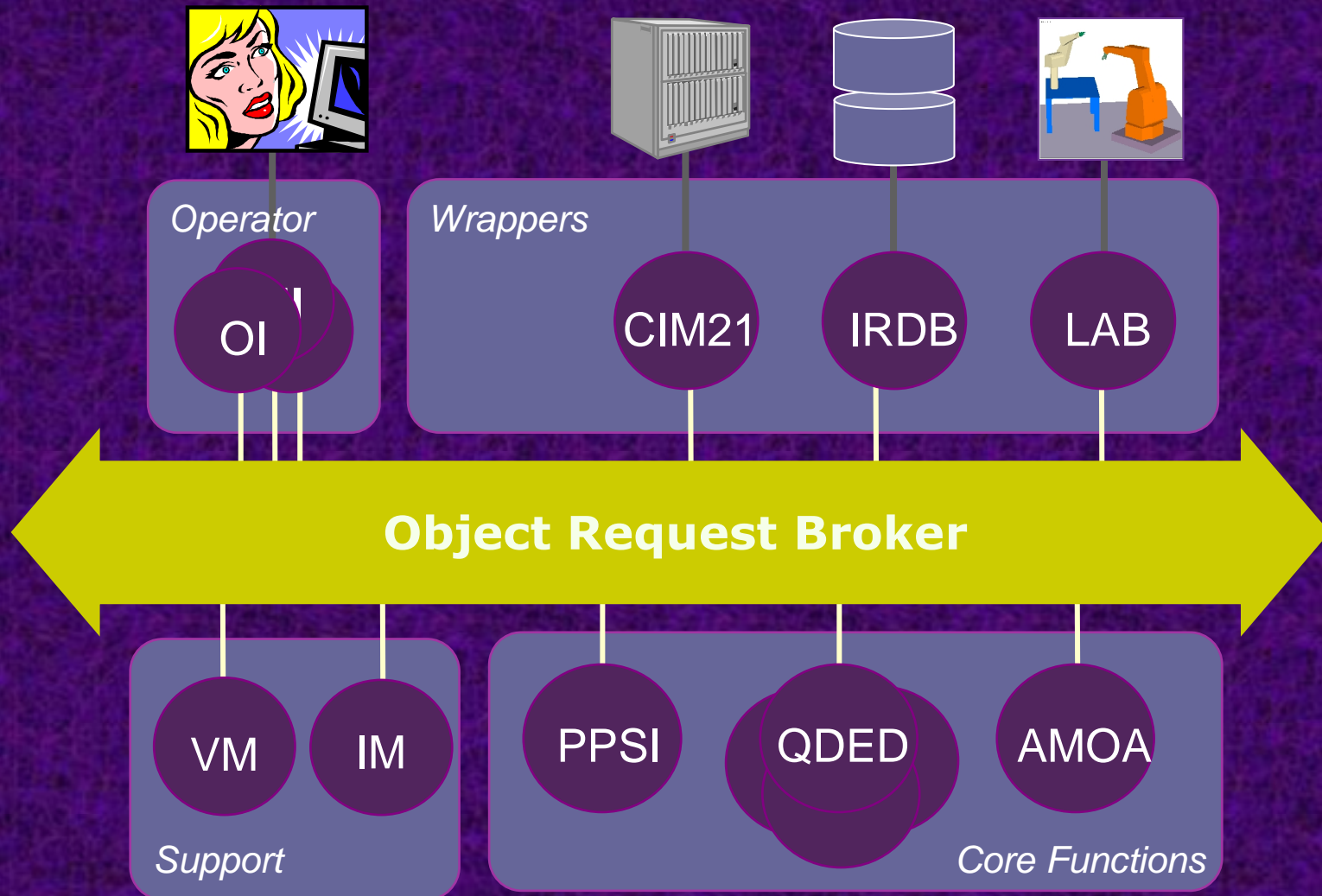
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- Remote Expert Control
- Expert system running on special processors controlling remote system
- Typically multi-layered control
  - Reactive real-time controller
  - Intelligent layers

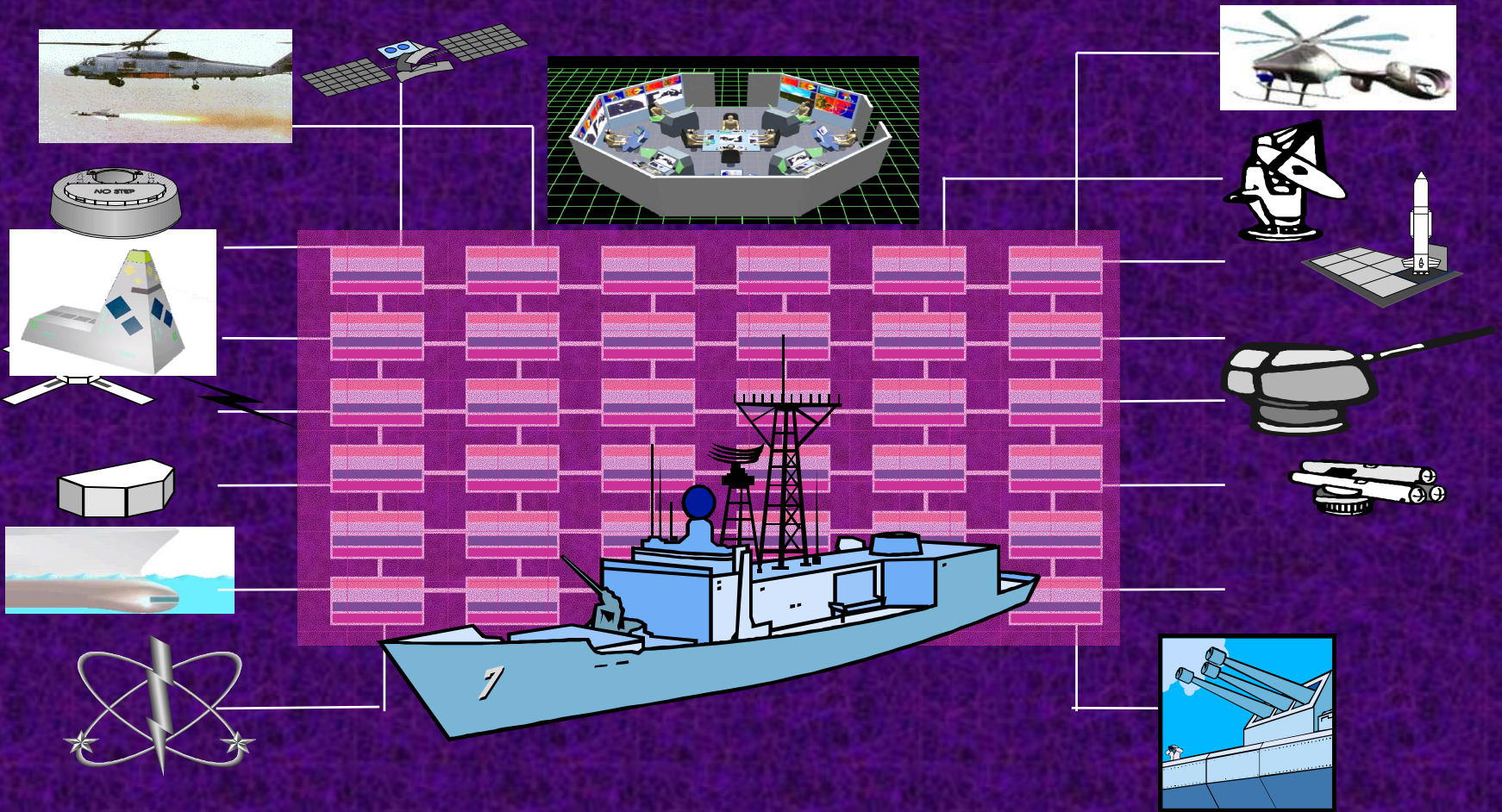




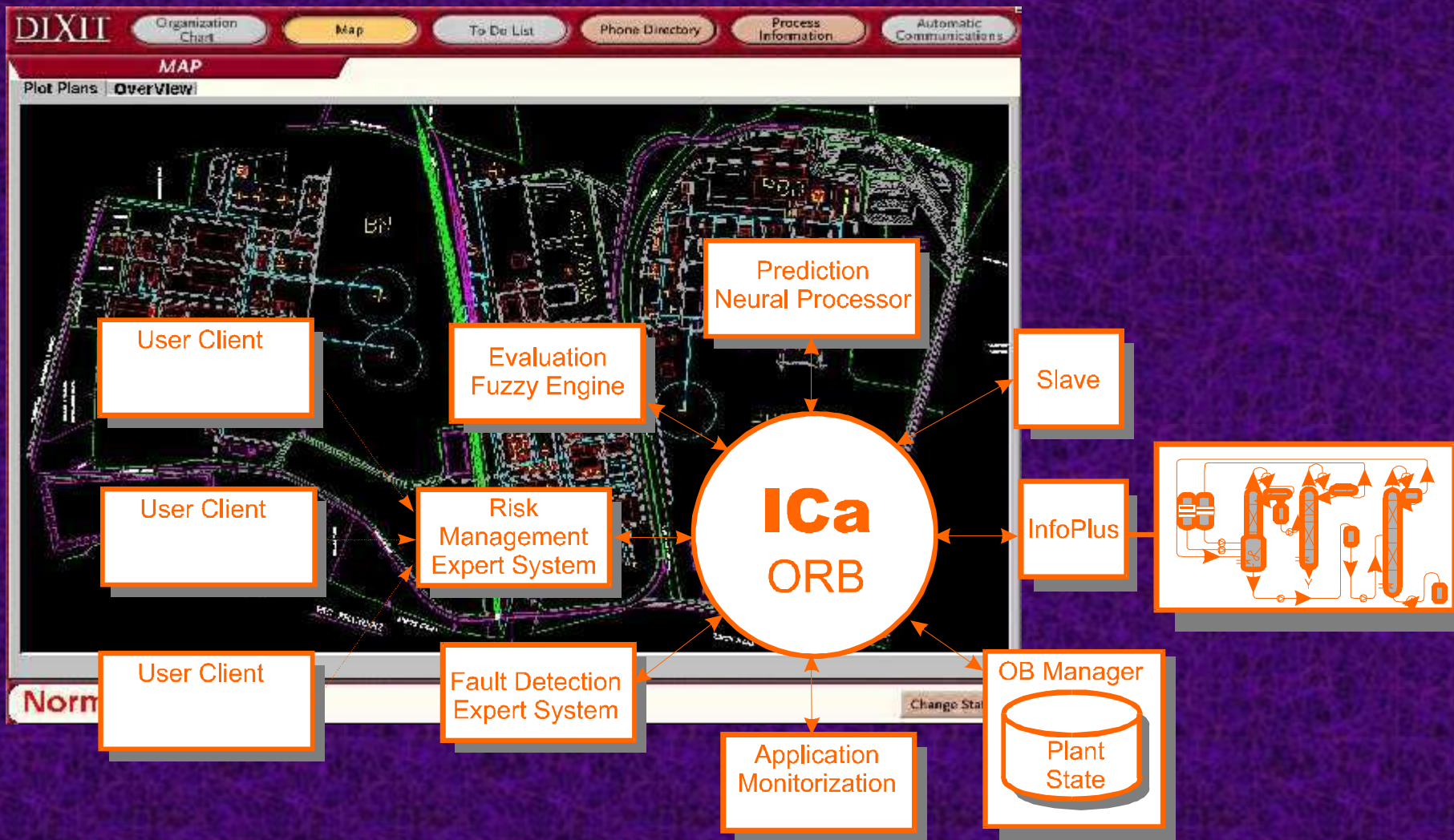
# Strategic Process Control



# Total-Ship Systems



# Plant-wide Risk Management

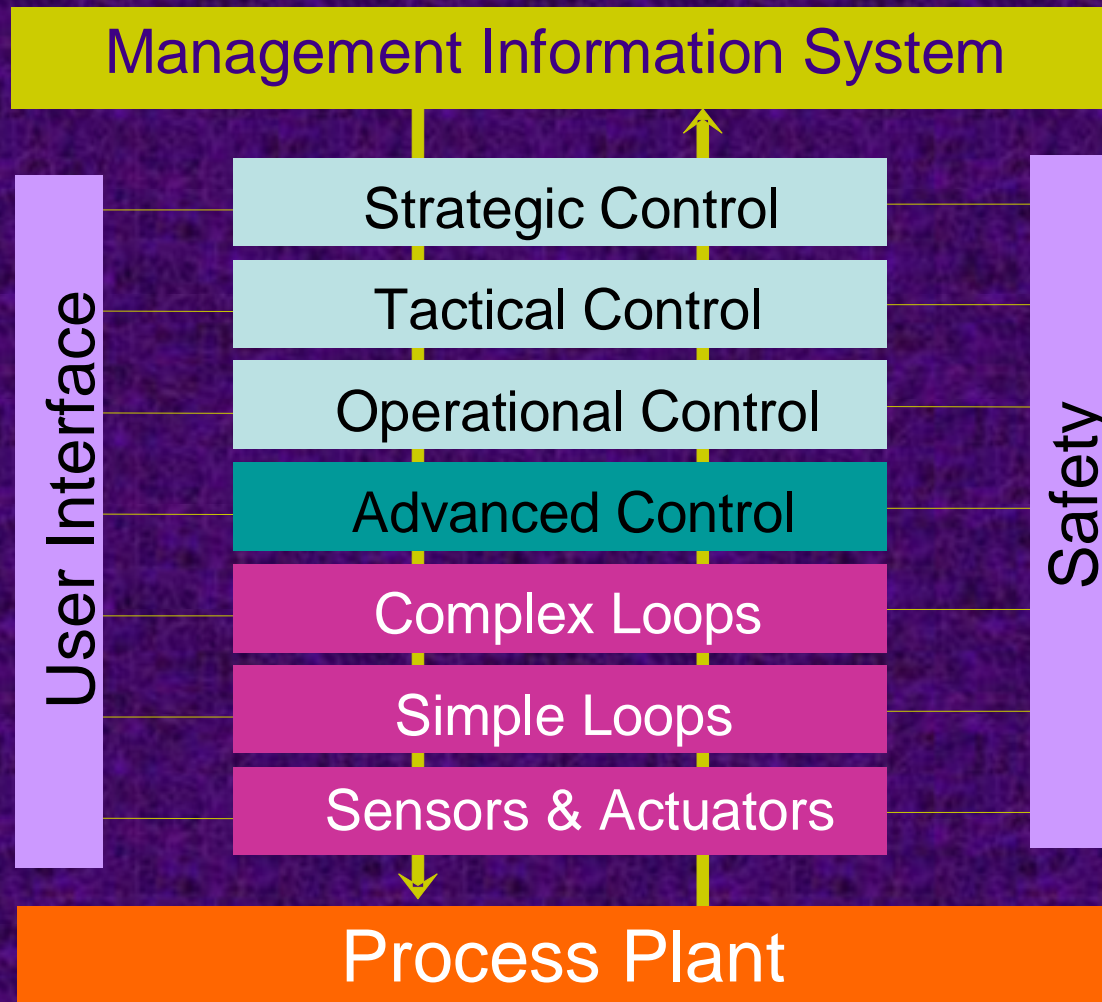


# Deeper Intelligence

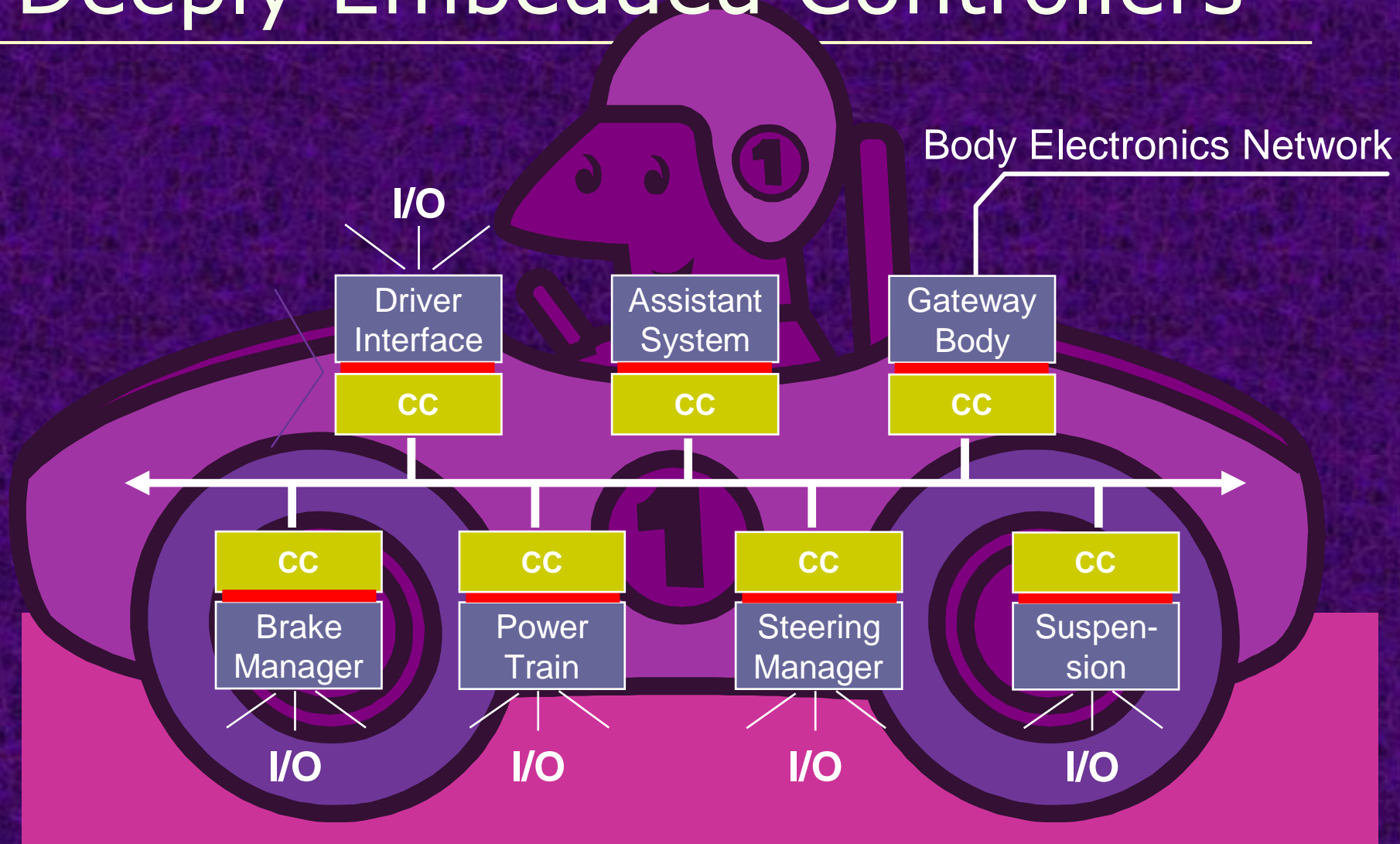


Increased intelligence levels in  
complex controllers

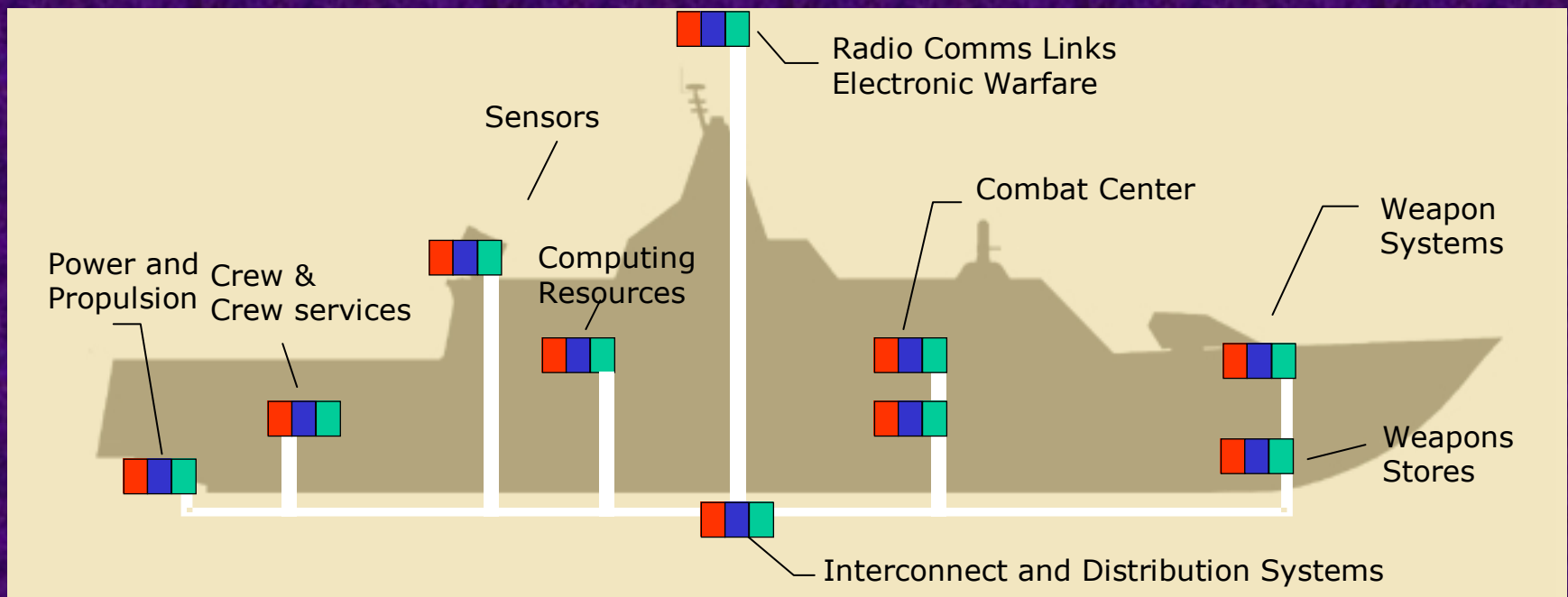
# Pervasive Intelligence



# Deeply Embedded Controllers



# Organic Systems of Objects



■ Physical Object ■ Logical Object (In Sys Mgmt Sense) ■ Network Connection

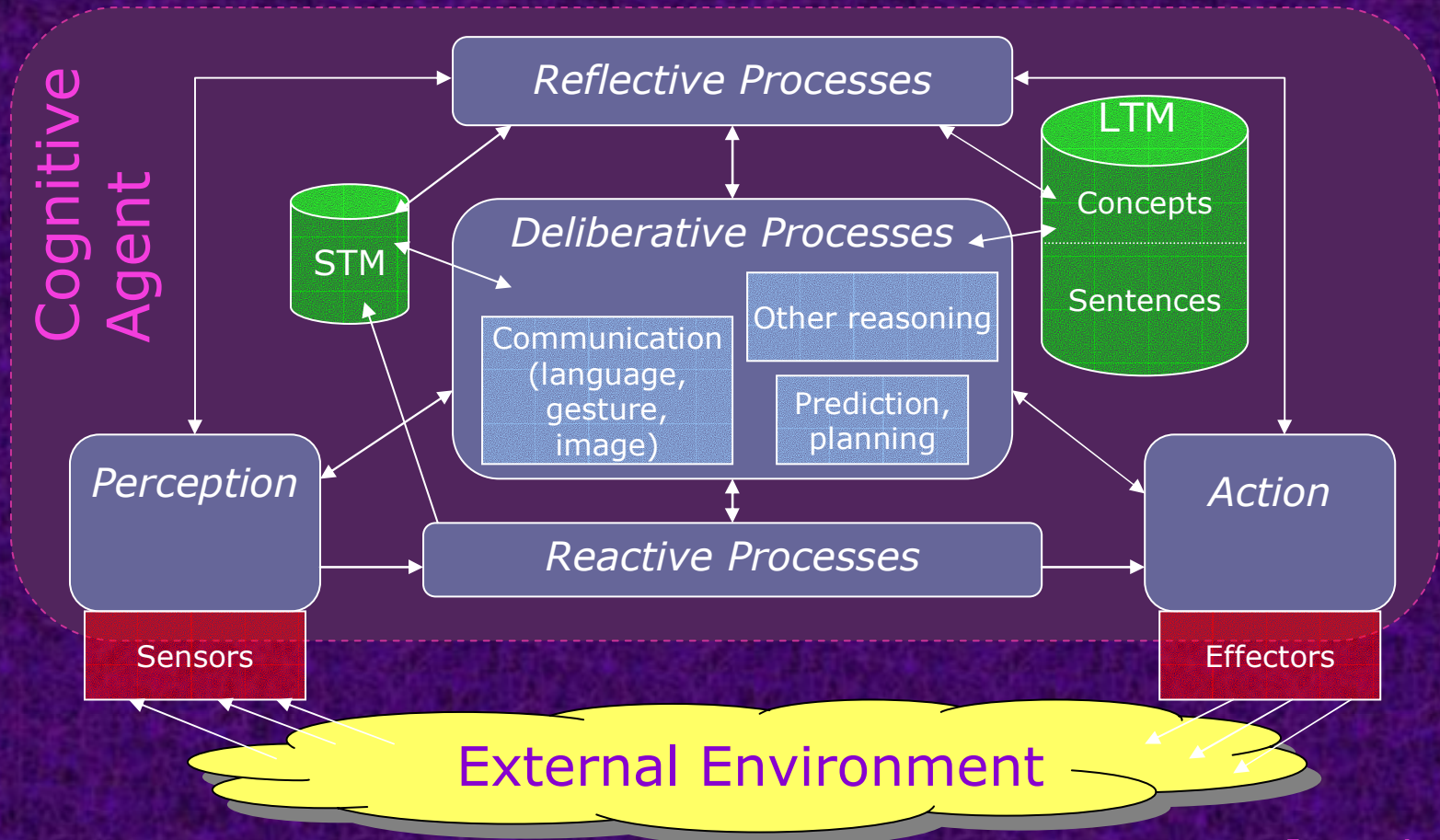
# Multiple agent architectures

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- Coarse grained architectures
  - BDI (Beliefs, desires, intentions)
  - GOMS (Goals, Operators, Methods, Selections)
  - Soar
- Fine grained architectures
  - Custom
  - FIPA.org
  - SIM\_Agent
  - OAA



# A Cognitive System Anatomy



[Brachman]

# Thrust in Cognitive systems

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- A cognitive system is one that
  - can **reason**, using substantial amounts of knowledge
  - can **learn** from its experience so that it performs better tomorrow than it did today
  - can **explain** itself and be told what to do
  - can **be aware** of its own capabilities and reflect on itself
  - can **respond robustly** to surprise

[DARPA]

- A cognitive system is one that
  - is physically instantiated or **embodied**
  - can perceive and **understand** (the semantics of information conveyed through their perceptual input)
  - Can **interact** with their environment
  - can **evolve** in order to achieve human-like performance in activities requiring context-(situation and task) specific knowledge

[IST]

# Beyond “Normal” Agents

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- Dependable control agents do have **requirements** well beyond what is considered “normal” intelligent function:
  - Real-time behavior
  - Embeddability
  - Evolvability
  - Upgradeability
  - Robustness

# Mechanisms for robustness

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- $H_2, H_\infty$  (robust control)
  - The system tolerates small displacement from design conditions
- Redundancy
  - Increase robustness up to a limit where the increase in dependability is less than the new induced risks
- Fault-tolerance
  - Copes with plant changes due to faults
- Reflection

# Robust design

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- “The suggestion that robust design is the **primary source of complexity** is motivated by the observation that for most biological and technological systems, the vast majority of components are present for robustness rather than for basic functionality of the organism or machine.”

[Reynolds]

# Two Notions of Complexity

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- “Complexity emerges in systems that are otherwise internally homogeneous and simple”.
  - **Self-organized criticality** (SOC) and the edge of chaos suggests that large-scale structure arises naturally and at no apparent cost through collective fluctuations in systems with generic interactions between individual agents.
- “Complexity is associated with intricately designed or highly evolved systems”.
  - **Highly optimized tolerance** (HOT) emphasizes the role of robustness to uncertainties in the environment as a driving force towards increasing complexity in biological evolution and engineering design.

# Scalable Intelligence



One-size fits-all intelligences

# Levels of Machines

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- ❑ The machine as a **tool**: The user can execute an action with the machine
- ❑ The machine as an **automaton**: The user (or condition) can initiate action sequences
- ❑ The machine as an **agent**: The user can request a context dependent function
- ❑ The machine as an **autonomous agent**: The machine executes context dependent actions as is deemed necessary by (heuristic) set of rules
- ❑ The machine as a **cognitive agent**: The machine understands and is aware; is able to execute tasks requiring real intelligence and thought

[Haikonen]



# Problems

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- ❑ Traditional control systems construction processes are based on **proven methods** that can lead to well-performing systems.
- ❑ Conventional control systems **modules** for distributed controllers can be built and are built this way.
- ❑ But there is a major problem when we address complex systems:
  - traditional engineering methods for high-assurance, distributed or intelligent control systems **do not scale well**.

# Scalable Control

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## □ Scalable

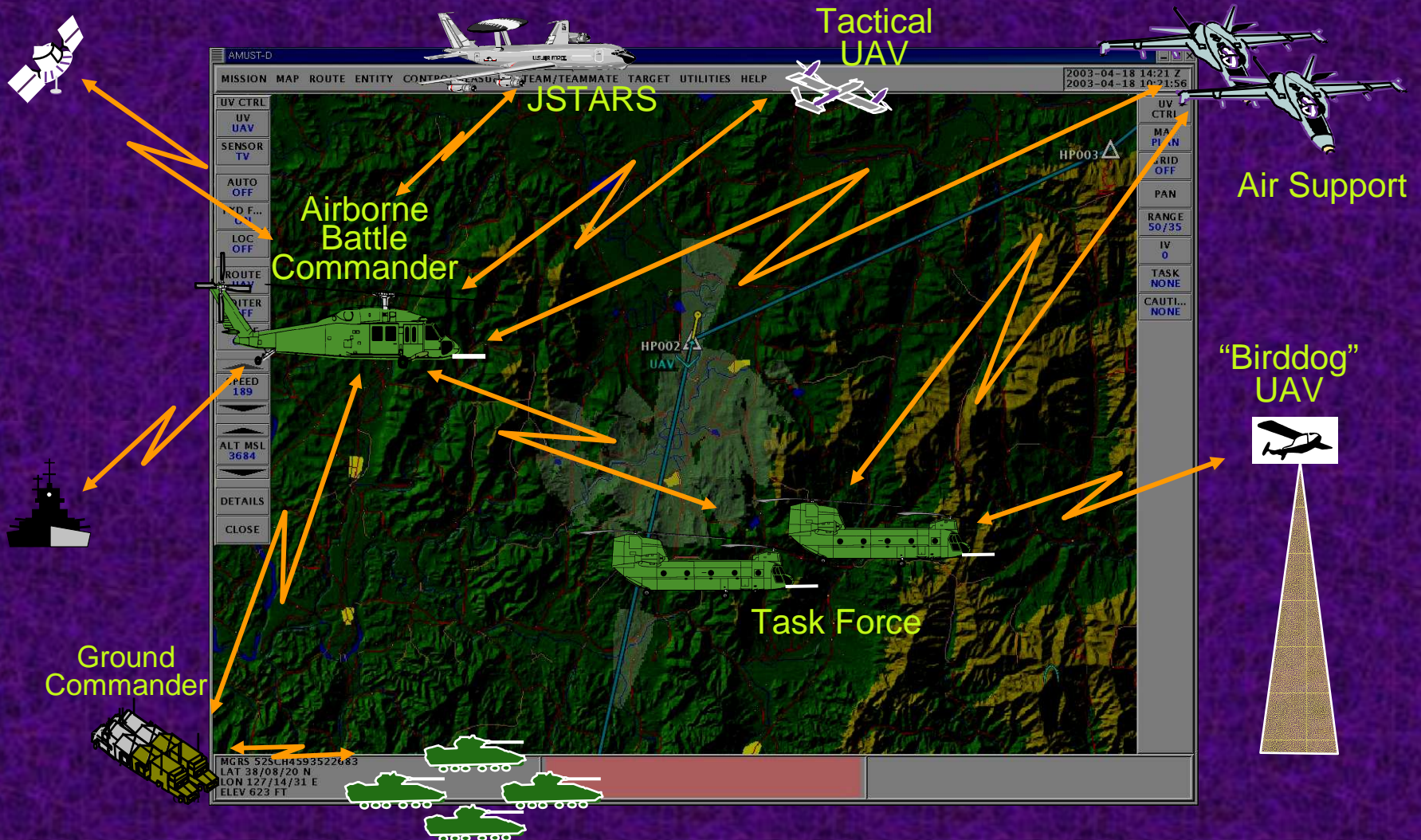
\Scal"able\, a. Capable of being scaled.

## □ Scaled to what ?

## □ Dimensions

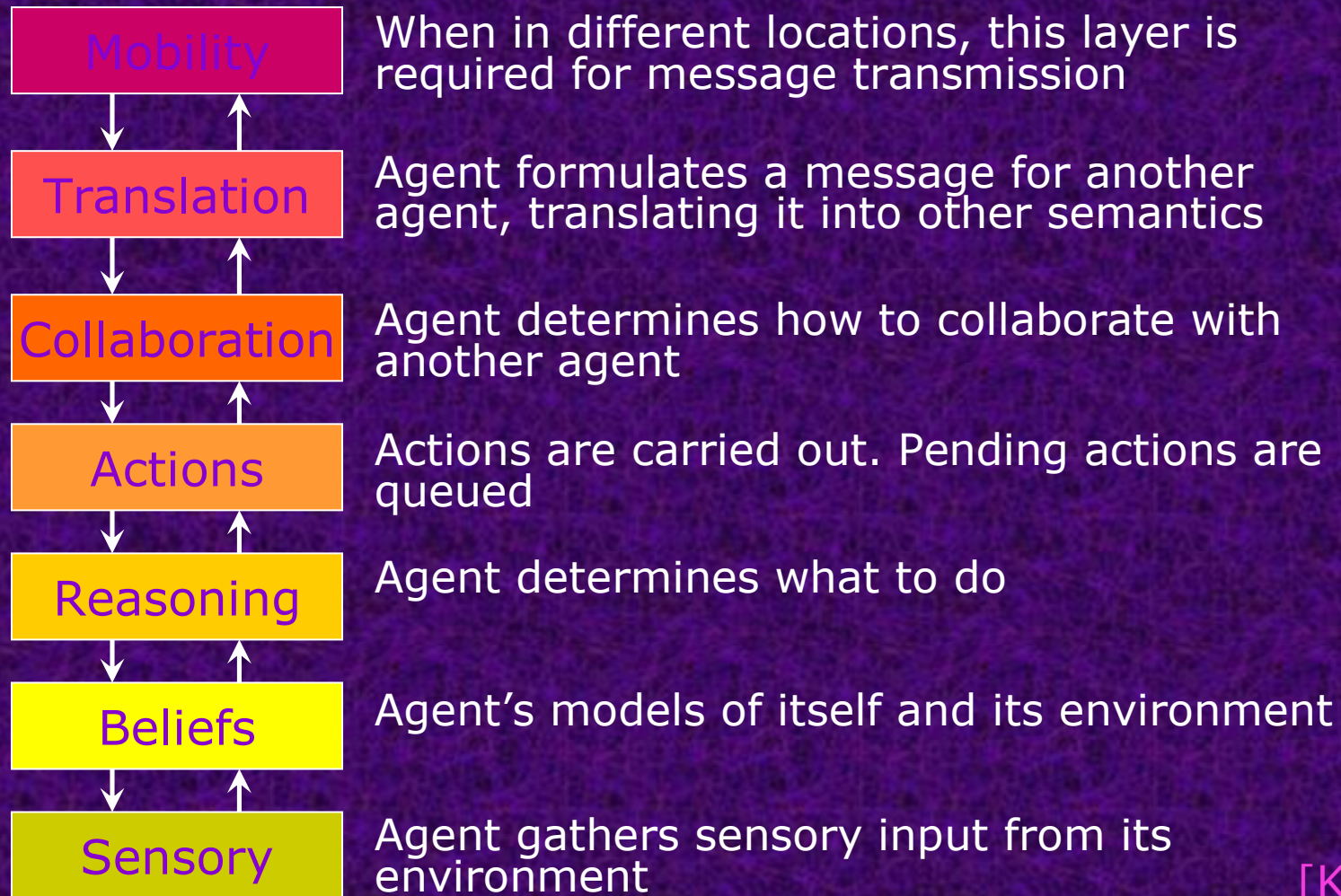
- **Space**: complex wide-area plants
- **Time**: multiple time-scale loops
- **Rationality**: levels of thought
- **Size**: down to embeddability

# Societies of Intelligences



# The Layered Agent Pattern

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[Kendall]

# Scaling Intelligence

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- Increasing intelligence levels to meet requirements
- Use of heterogeneous technologies
- Use of control encapsulation
  - PID
  - Fuzzy
  - Rule-based Learning
- Now touching the frontiers of being-human

# Integrated Control Architecture

- Real-time Middleware
- Generic agents
  - Architectural
    - Core, RT, FT, BDI
  - Task
    - PlantState, Simulator, etc.
  - Interface
    - Wrappers, Probes, etc.
- Design Patterns

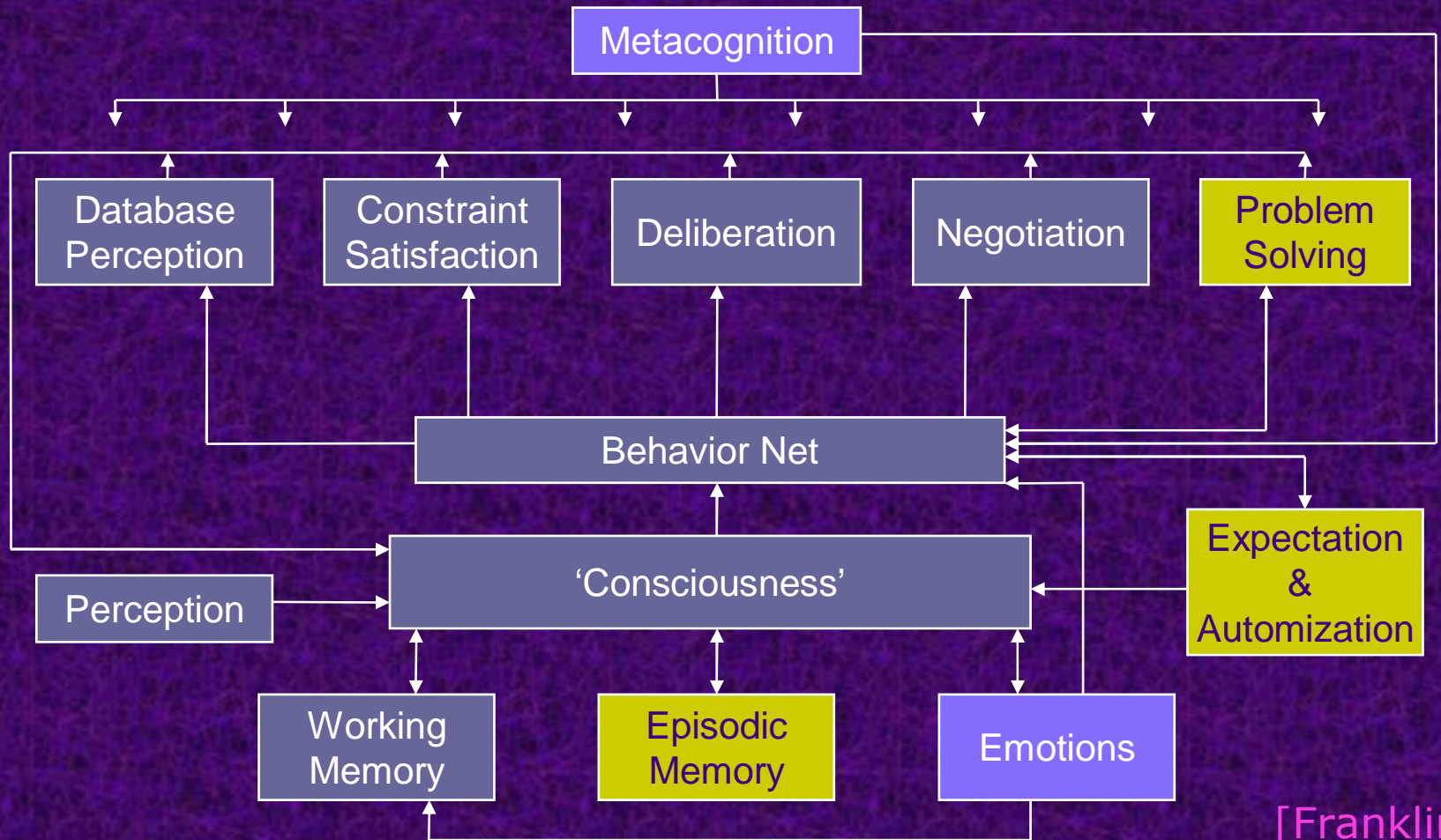


# Adding Reflection

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- ❑ Incorporates structures representing (aspects of) itself
- ❑ Go beyond what is called fault-tolerant control
- ❑ The self representation is **causally connected** to the aspects of the system it represents:
  - The system has always an accurate representation of itself
  - The status and computation of the system are always in compliance with this representation

# IDA's Architecture



[Franklin]



# Autonomic Systems

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## (as IBM sees them)

- ❑ Adapts to changes in its environment
- ❑ Strives to improve its performance
- ❑ Heals when it is damaged
- ❑ Defends itself against attackers
- ❑ Exchanges resources with unfamiliar systems
- ❑ Communicates through open standards
- ❑ Anticipates users' actions
  
- ❑ Possesses a **sense of self**

[SciAm May 06, 2002]

# Hot words

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- Meaning
- Value
- Awareness
- Consciousness
- Self
- Emotion
- Imagination
- Qualia
- Wisdom

# The question ?

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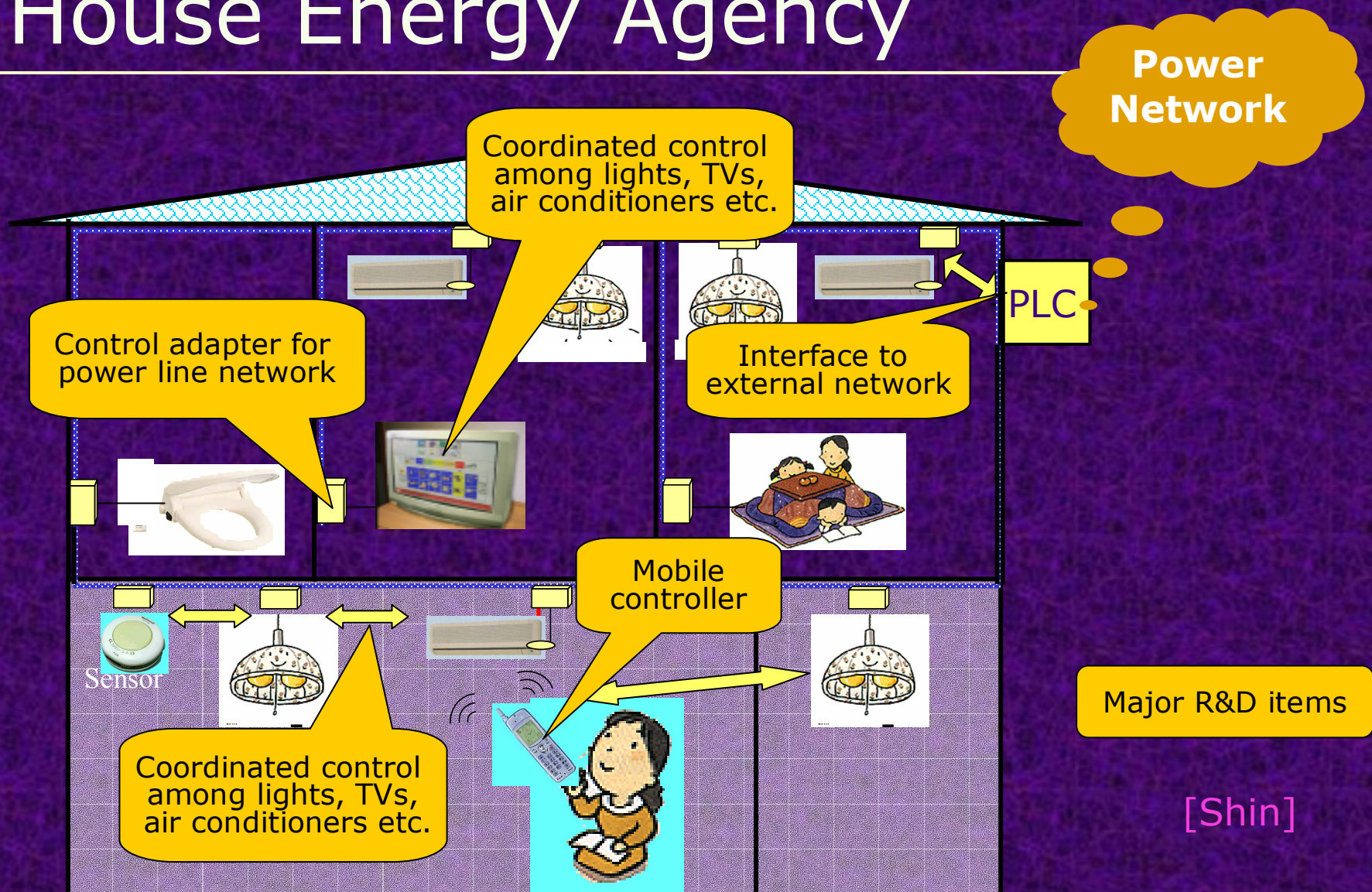
- Is there any agent architecture that could be scaled across dimensions and still guarantee critical control properties?

# Scalable Distributed Intelligence



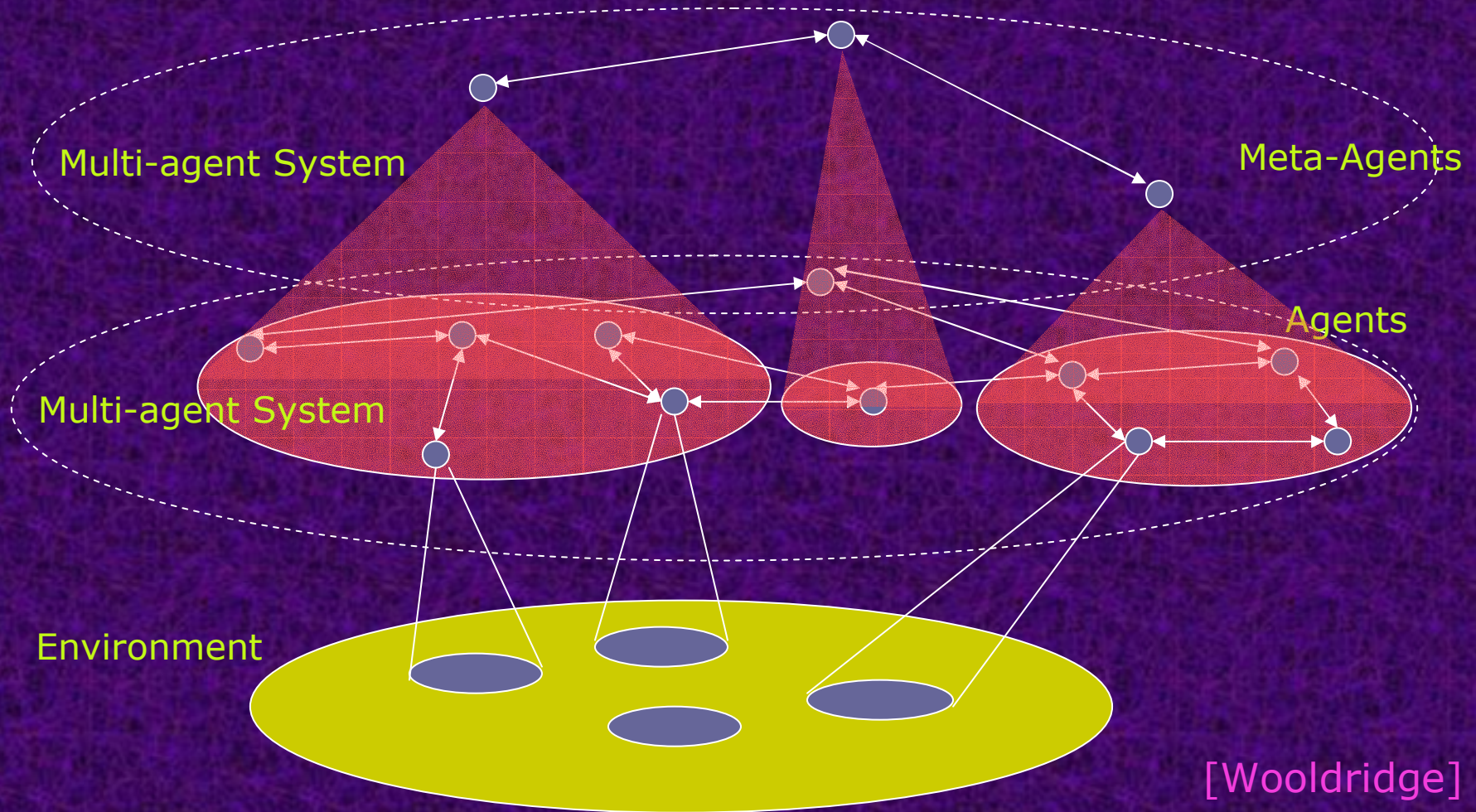
This is the wild speculation part

# House Energy Agency



[Shin]

# Agents of Agents



[Wooldridge]

# Modules and protocols

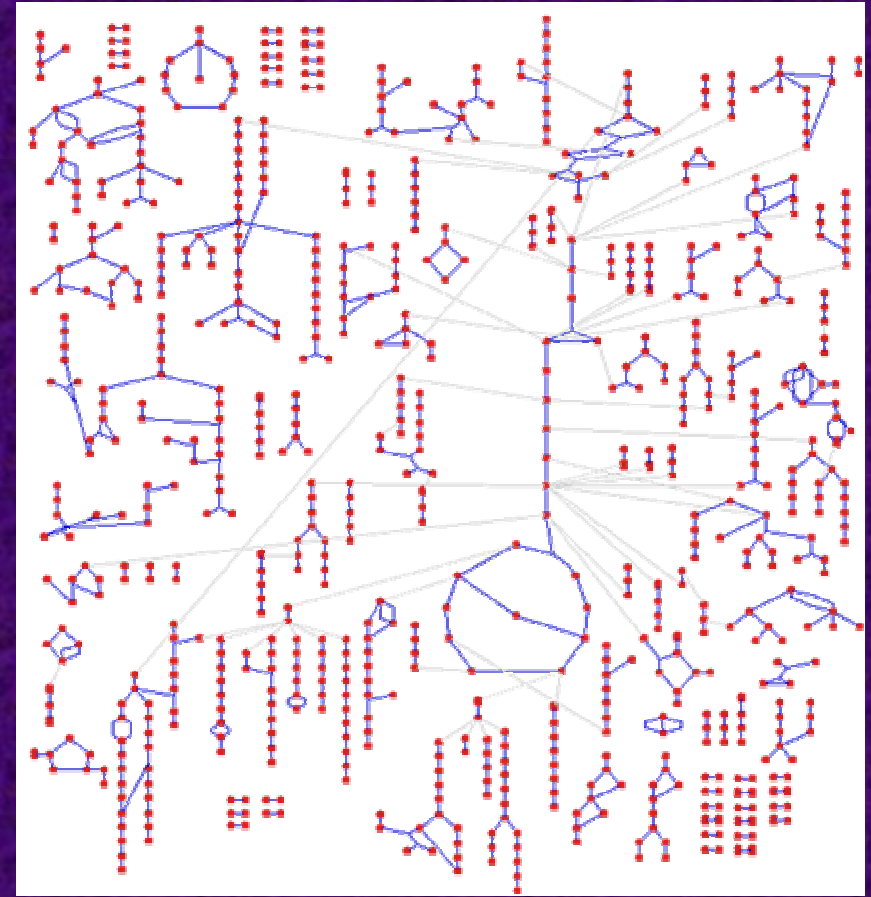
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- **Protocols** and **modules** are complementary (dual) notions
- Primitive technologies
  - Modules are more important than protocols
- Advanced technologies
  - Protocols are at least as important
- Even bacteria are “advanced technology”

[Doyle]

# Biochemical Network

## □ E. Coli Metabolism

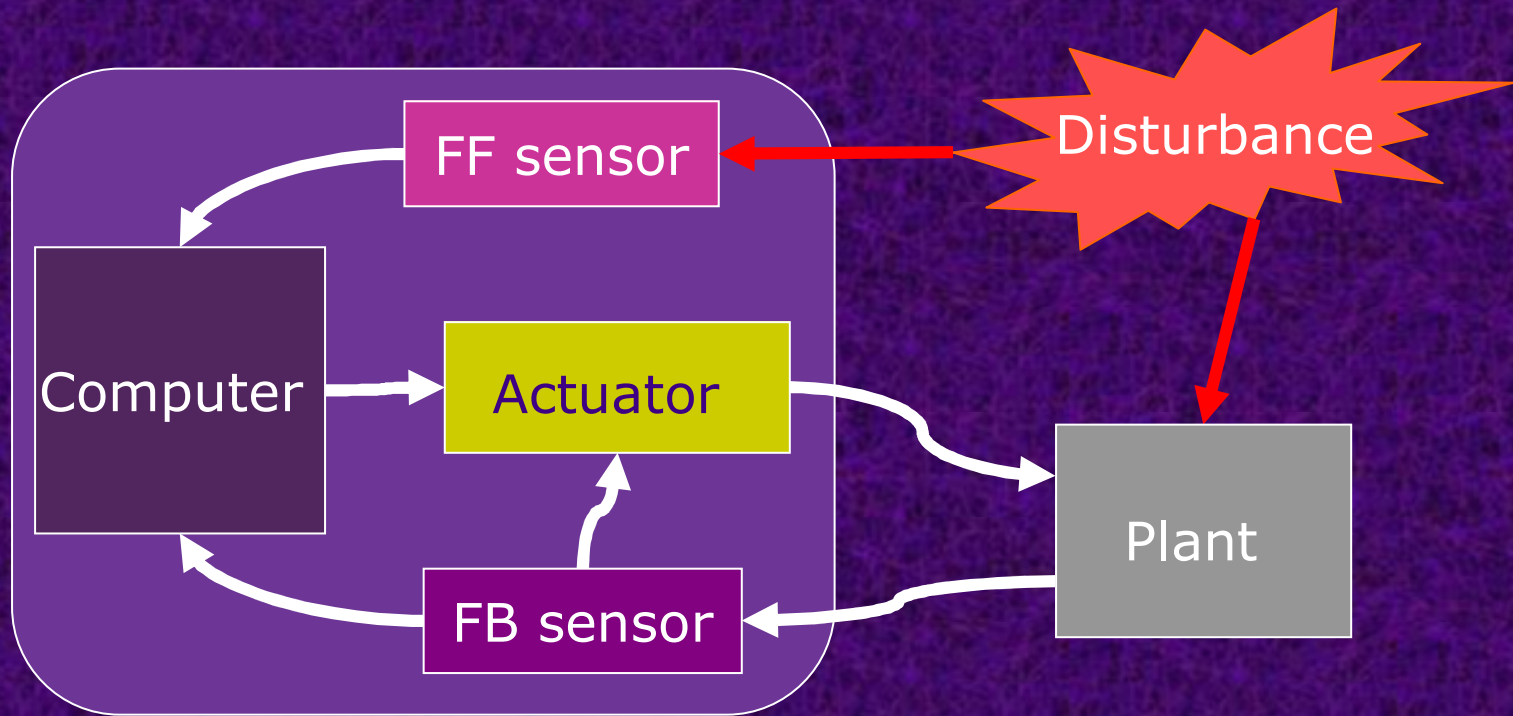


[Arkin]



# A “universal” architecture?

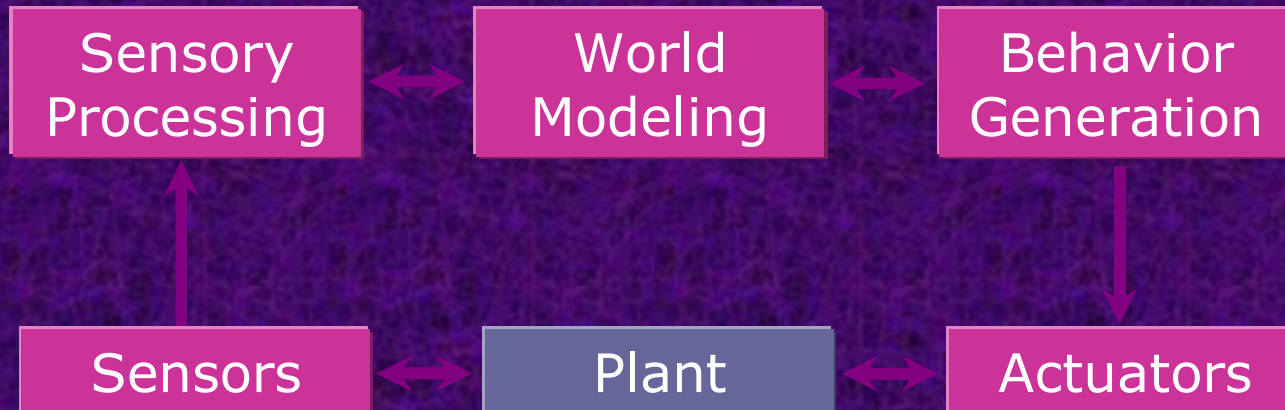
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[Doyle]

# A “universal” architecture?

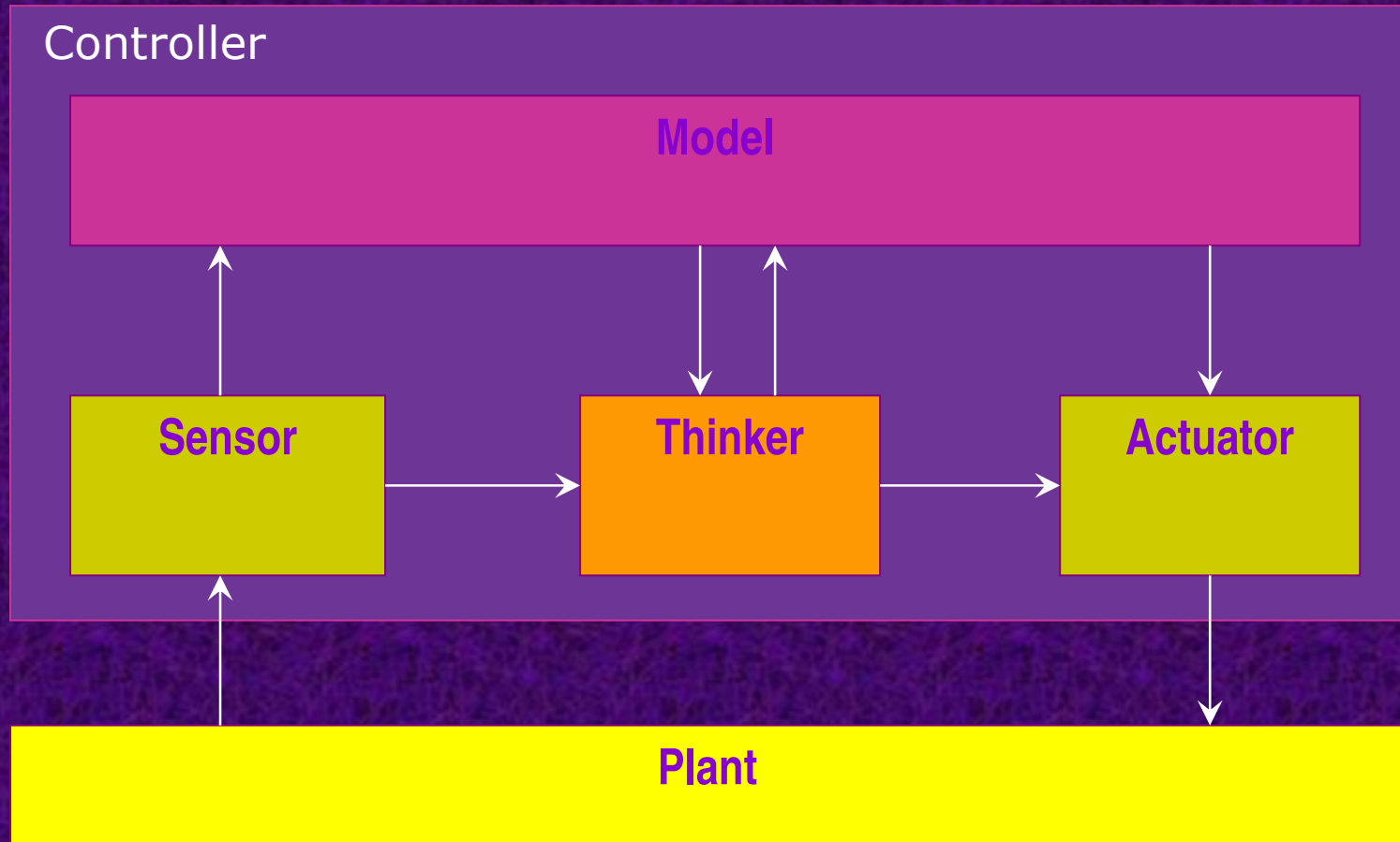
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[Meystel]

# A “universal” architecture?

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# Is there a core architecture ?

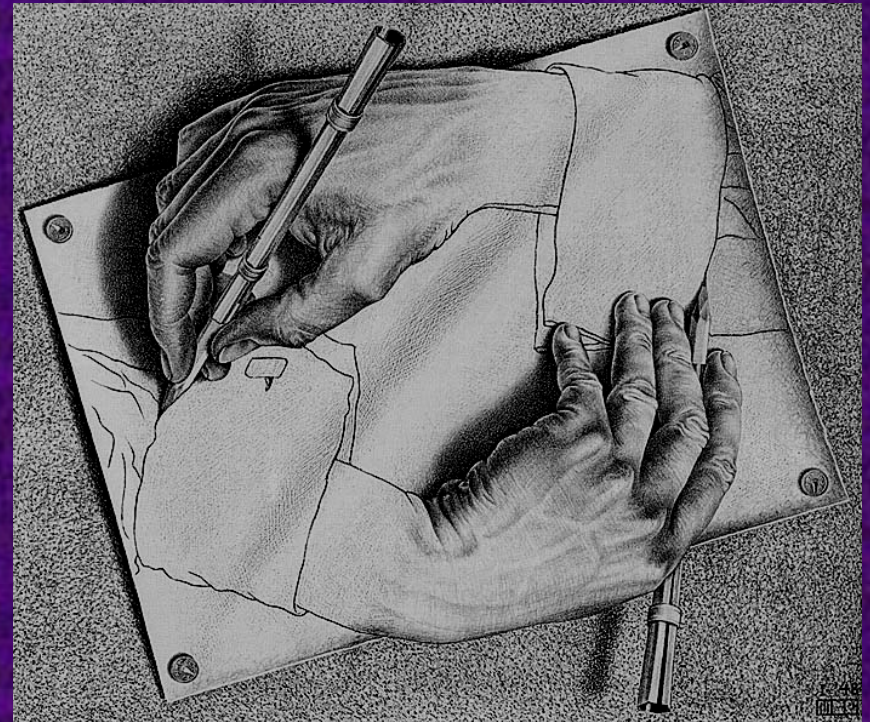
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- Feedback-feedforward loop?
  - Meystel-Albus' elementary loop of functioning?
  - Doyle's bowties?
  - Autopoietic networks?
  - Interactivist-constructivist systems?
- 
- Why is it **necessary** ?
    - Homogeneity
    - Emergence

# The Autopoietic Mind?

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- The difference between code and data is contextual not ontological
- Control systems that generate themselves as products



# Theoretical foundations

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- Control theory
- Dynamical systems
- Computational complexity
- Information theory
- Statistical physics
- Cognitive science
- Neuroscience  
(ground truth)

[Doyle+]

Largely fragmented  
within isolated  
technical disciplines.

Unified theory would  
be both intellectually  
satisfying and of  
enormous practical  
value.

# Summary

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- ❑ **Mind** is a **multiresolutional phenomenon** of model-based adaptive-predictive control
- ❑ Minds generate and use **dynamic models of reality** including their bodies
- ❑ Minds can be **scaled and aggregated** to form new minds
- ❑ At any resolution level, meaning generation generates **awareness**
- ❑ At any resolution level, mind reflection generates **consciousness**
- ❑ Scalable controllers should **imitate?** this basic design

# Thanks



Questions ?



## Abstract

As plants grow in complexity and control requirements grow in difficulty, there is an increased need for intelligence in control systems. Intelligence, traditionally a duty for central brains, has to be available where it is needed and that means, today, everywhere. This is achieved putting computing power and software in many places far from central brains. Intelligence is getting ubiquitous in control systems. Controllers are becoming societies.

Intelligence is distributed across the plants for several reasons (both needs and enablers): availability of suitable embeddable processors, timing requirements that forbid communication due to latencies, need of increased levels of performance that is achieved through parallelism, simplification of construction and maintenance tasks through modularity, reduction of cost and time-to-market by means of component-based reuse, etc.

Traditional control systems construction processes are based on proven methods that can lead to well-performing systems. Conventional control systems modules for distributed controllers can be built and are built this way. But there is a major problem when we address complex systems: traditional engineering methods for high-assurance, distributed or intelligent control systems do not scale well. High performance plant-wide distributed controllers are getting so complex that conventional engineering methodologies fail to achieve the degree of predictability needed. Obviously one of the reasons is that the tools used by the control engineers are not suitable for this task; but to my understanding the most critical problem is that conventional control architectures do not provide the degree of scalability of intelligence that is needed.

Much work has been done in complex controller understanding (see for example the work in learning control, real-time artificial intelligence, reflective controllers, multiresolutional hierarchical control or emergence in control agent societies) but this work does not properly address the issue of scalability.

In this talk I will address the current state of the art in the implementation of complex controllers with distributed intelligence; will describe the basic technologies, the trends and will propose research topics for future pervasive, highly intelligent, scalable controllers.