

Acting is the Purpose of Planning

or

The Actor's view of Deliberation

Malik Ghallab

A change of focus

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- ▶ Not a novel research path

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- ▶ Focus Automated Planning research on the *actor's perspective*

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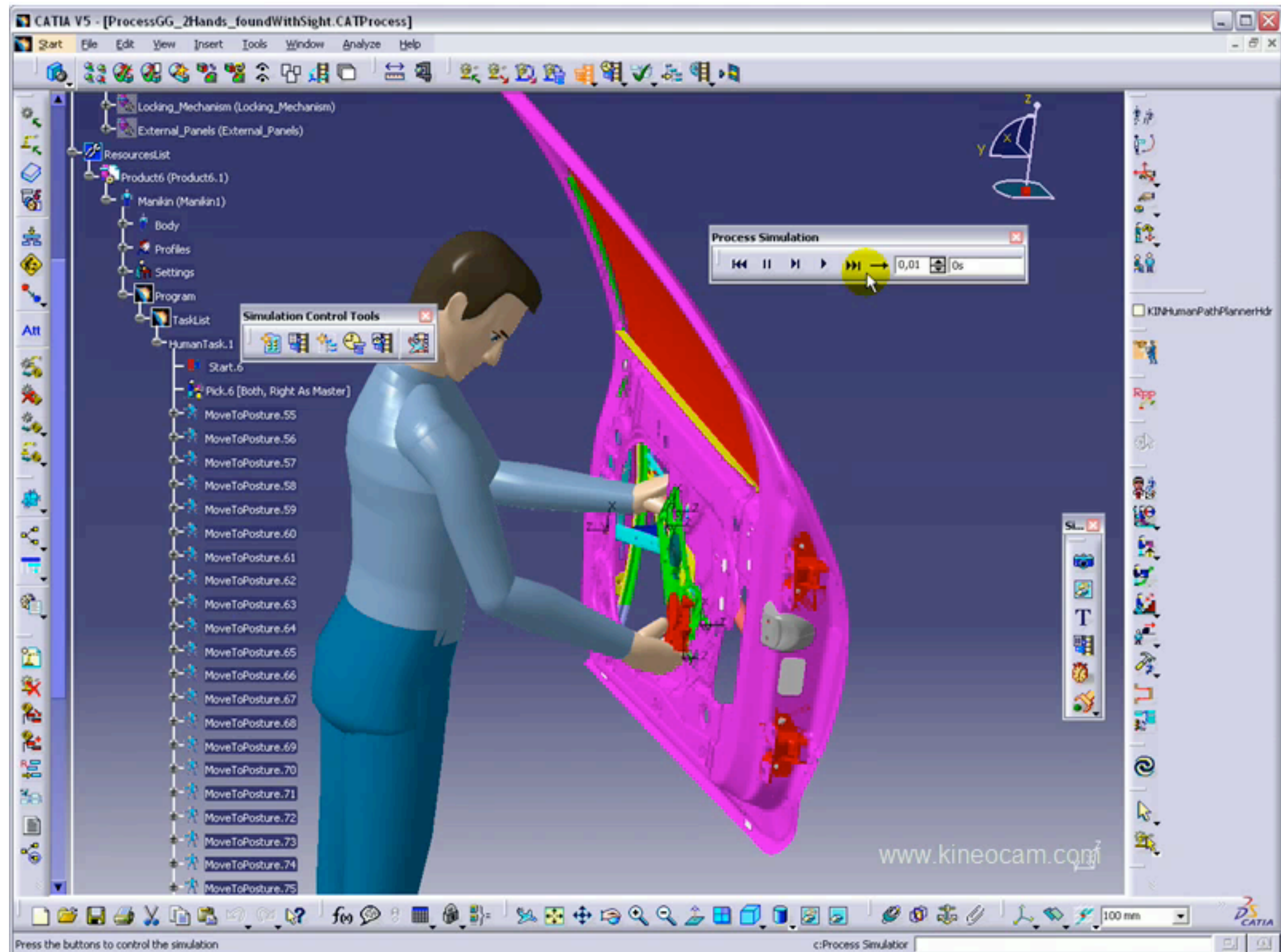
A Survey on Deliberation Functions for Autonomous Robots

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 - Collaboration with Felix Ingrand
A Survey on Deliberation Functions for Autonomous Robots
 - Collaboration with Dana Nau and Paolo Traverso
A Position Paper on Automated Planning and Acting

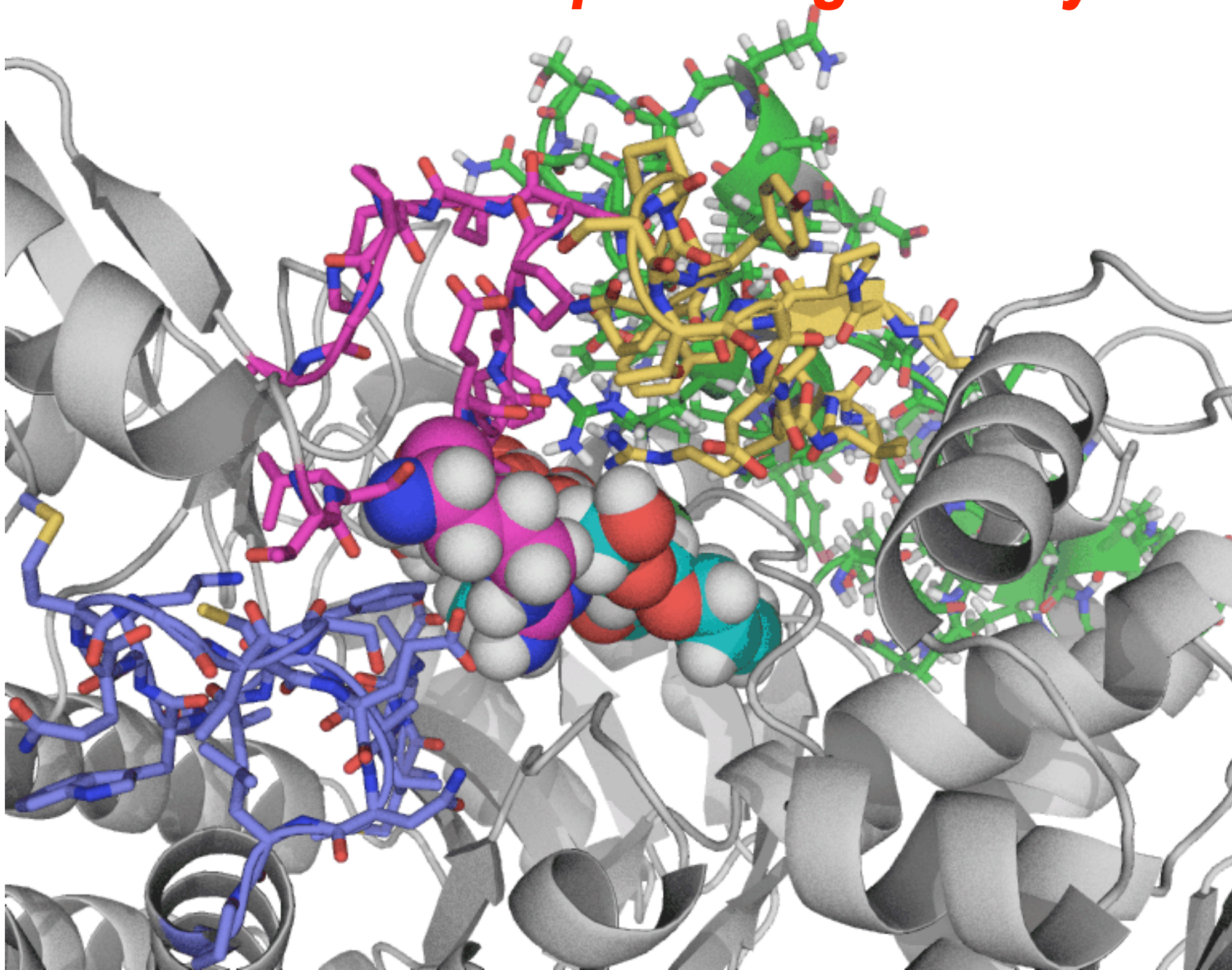
- ▶ Planning is a deliberation function
- ▶ Main purpose of planning is acting
 - Planning is valuable for other uses than acting
 - *Design* of tools, assembly, molecules permitting actions
 - Video games

Motion planning in car design



[Kineo]

Motion planning in enzyme design



[J.Cortes, LAAS]

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Deliberate action:

- Intentionally chosen and planned to achieve some objective
 - Pursued for the accomplishment of the robot's task
- ▶ Automated Planning research *does not*
focus on actual challenges of deliberate action

Critical view of current main stream

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 - Numerous “*Planners*”

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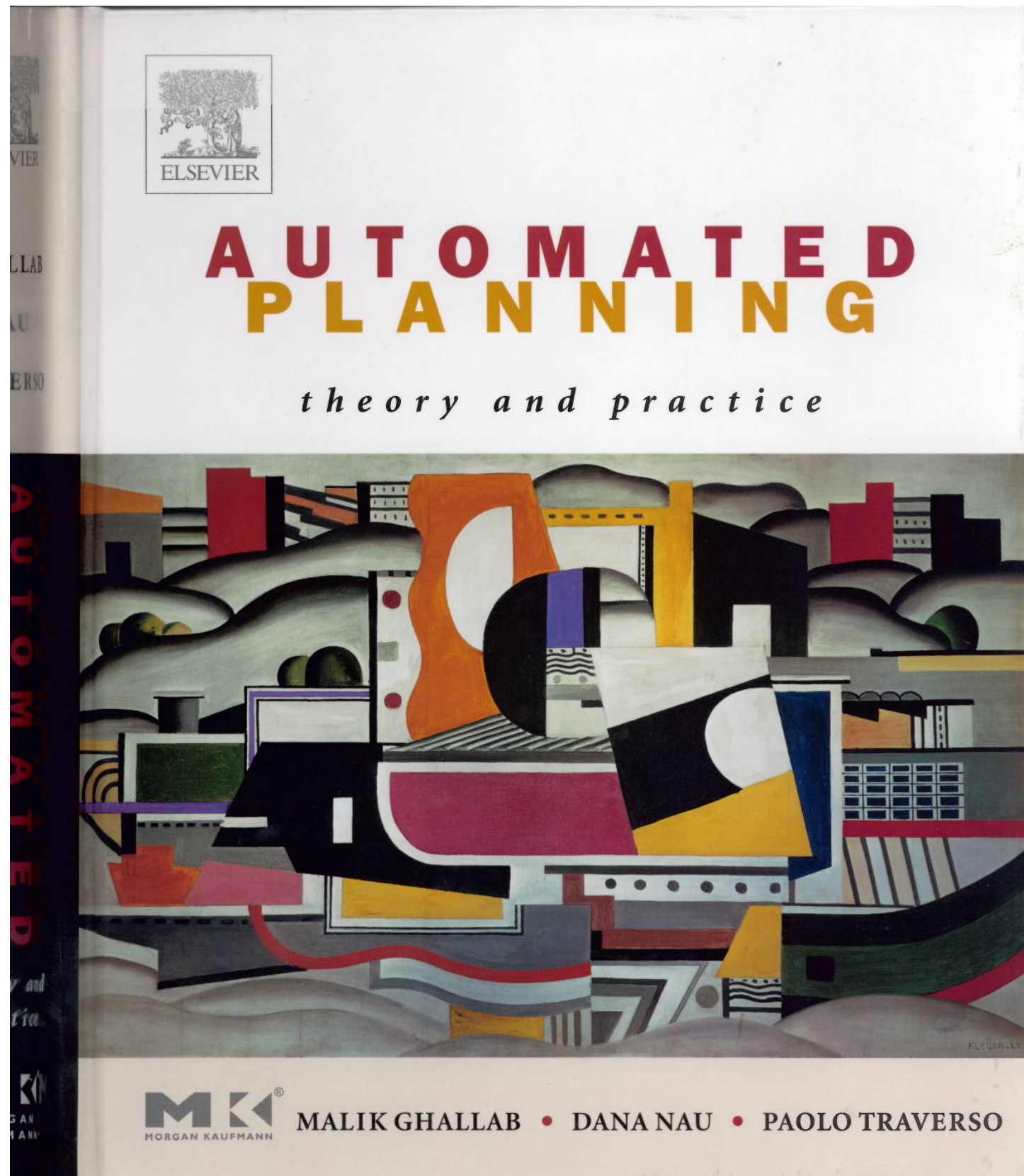
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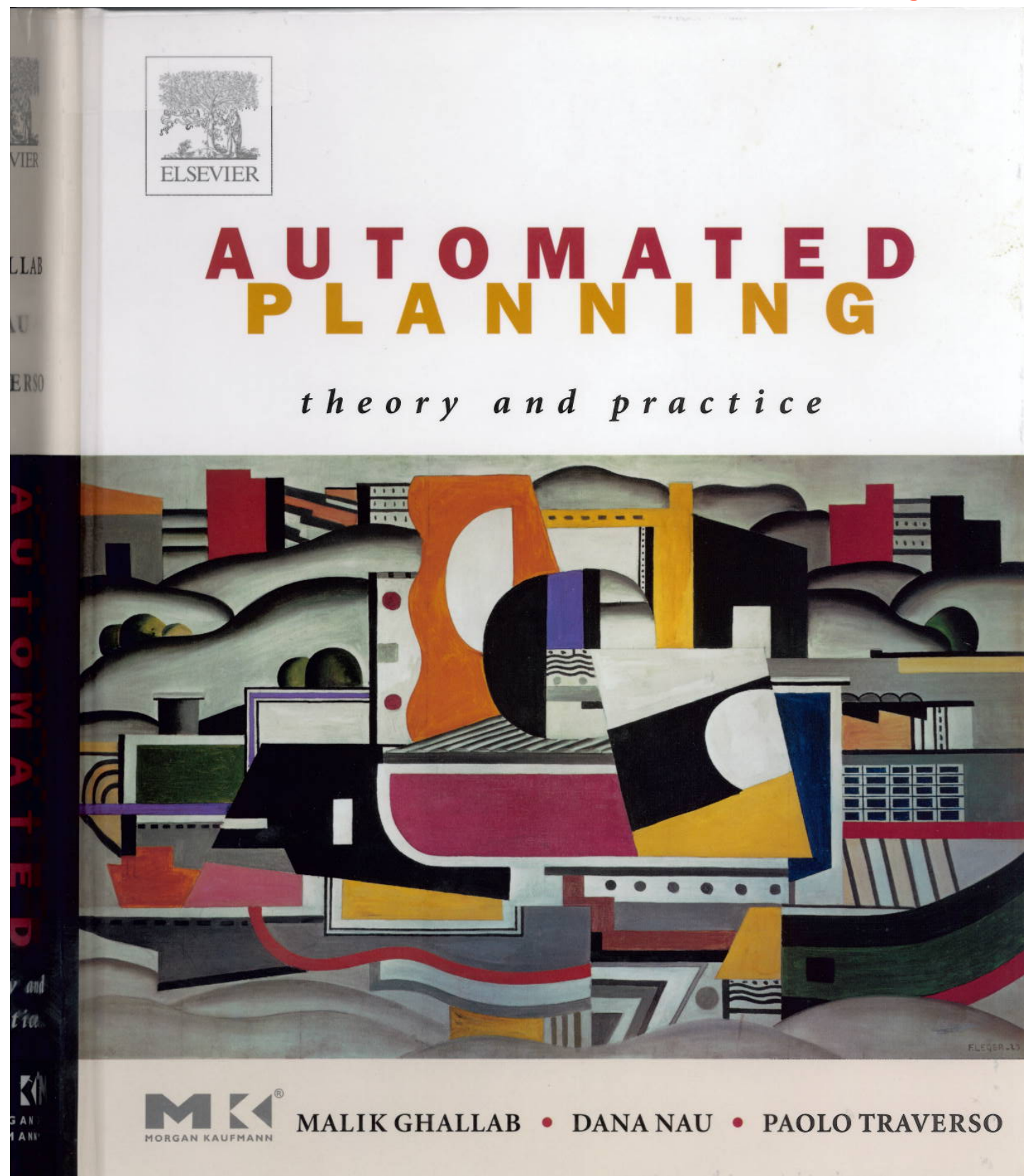
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 - Acting problems: insufficiently investigated

A typical textbook of the field

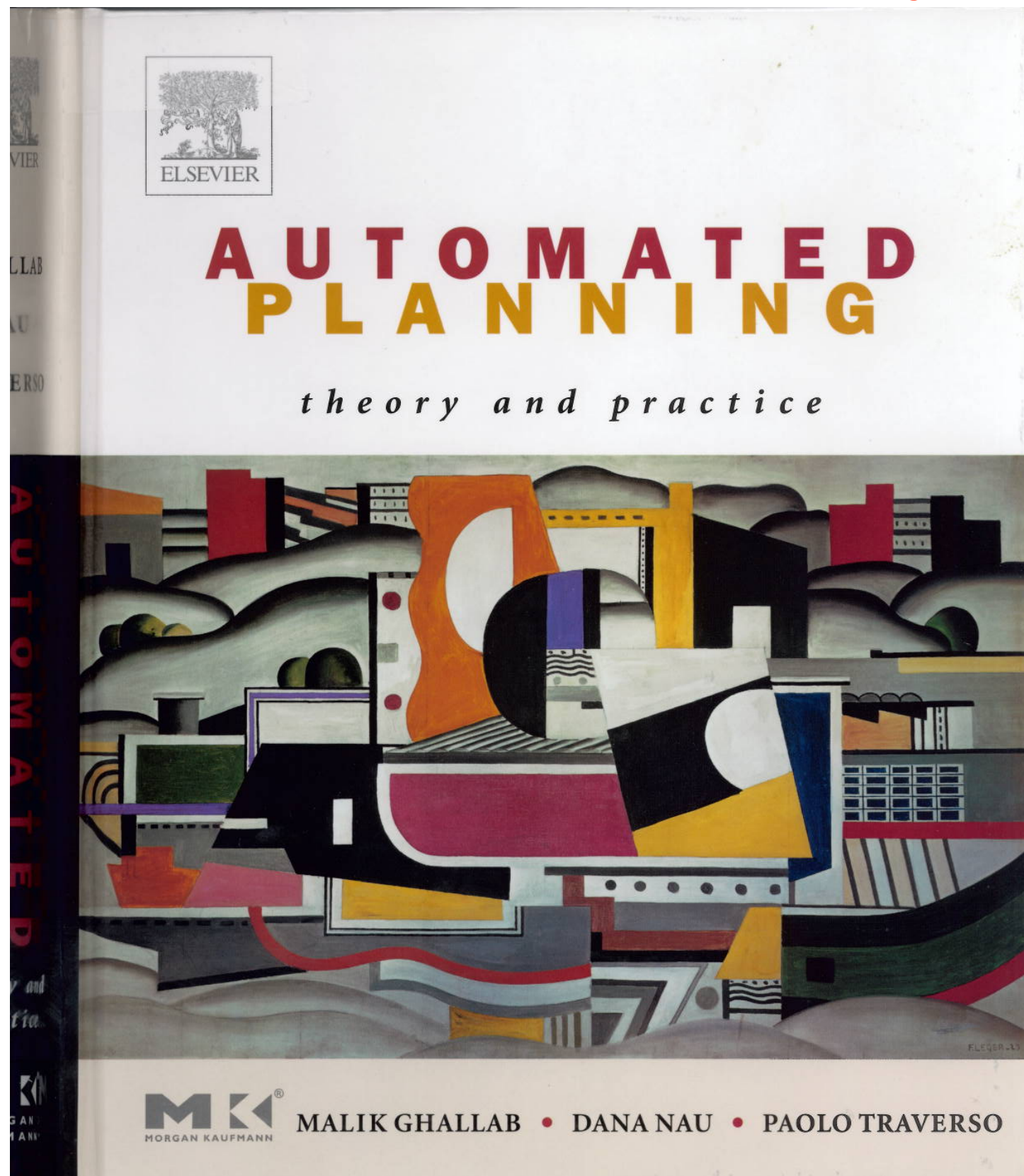


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► Planing and Acting:

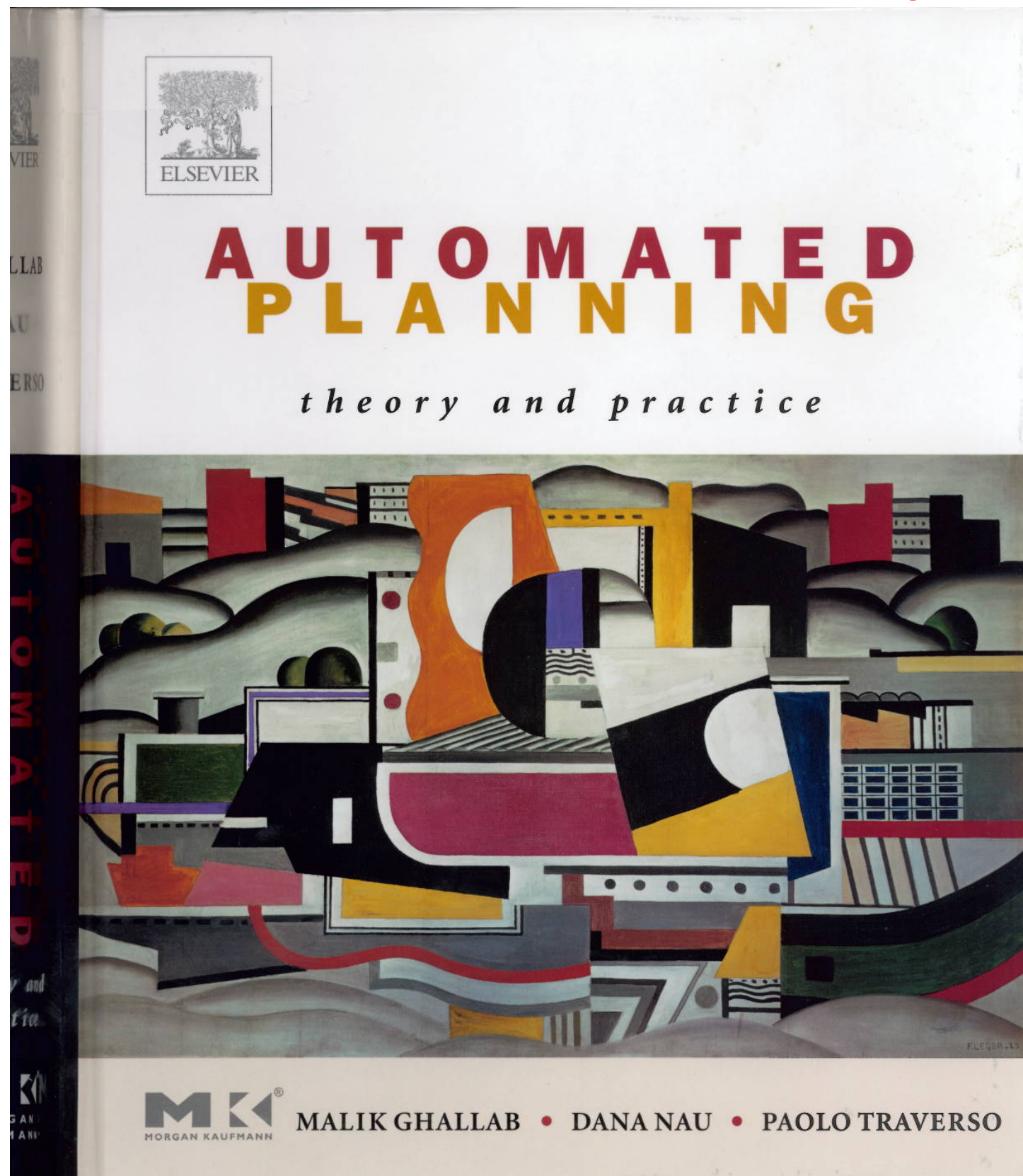


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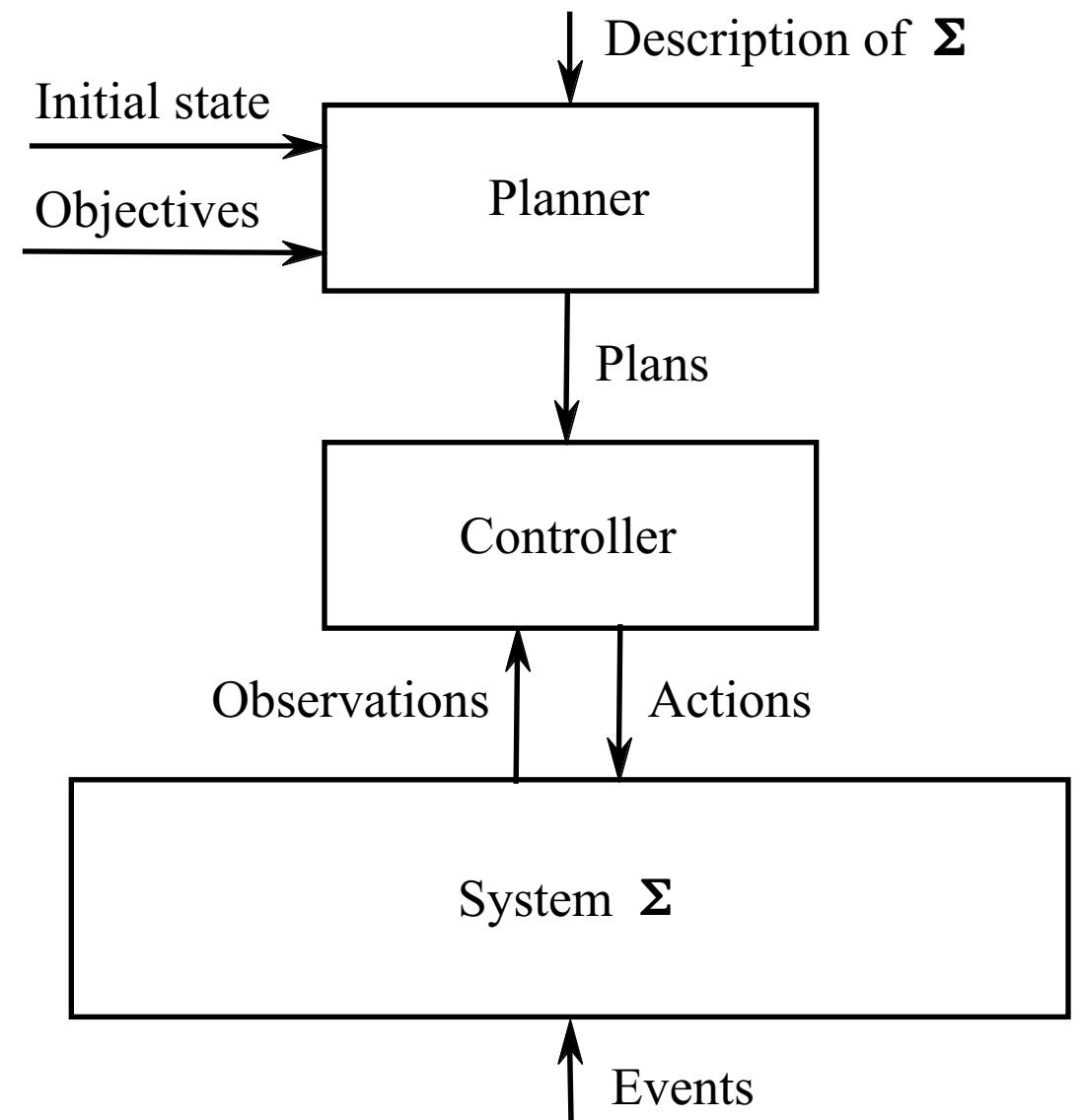
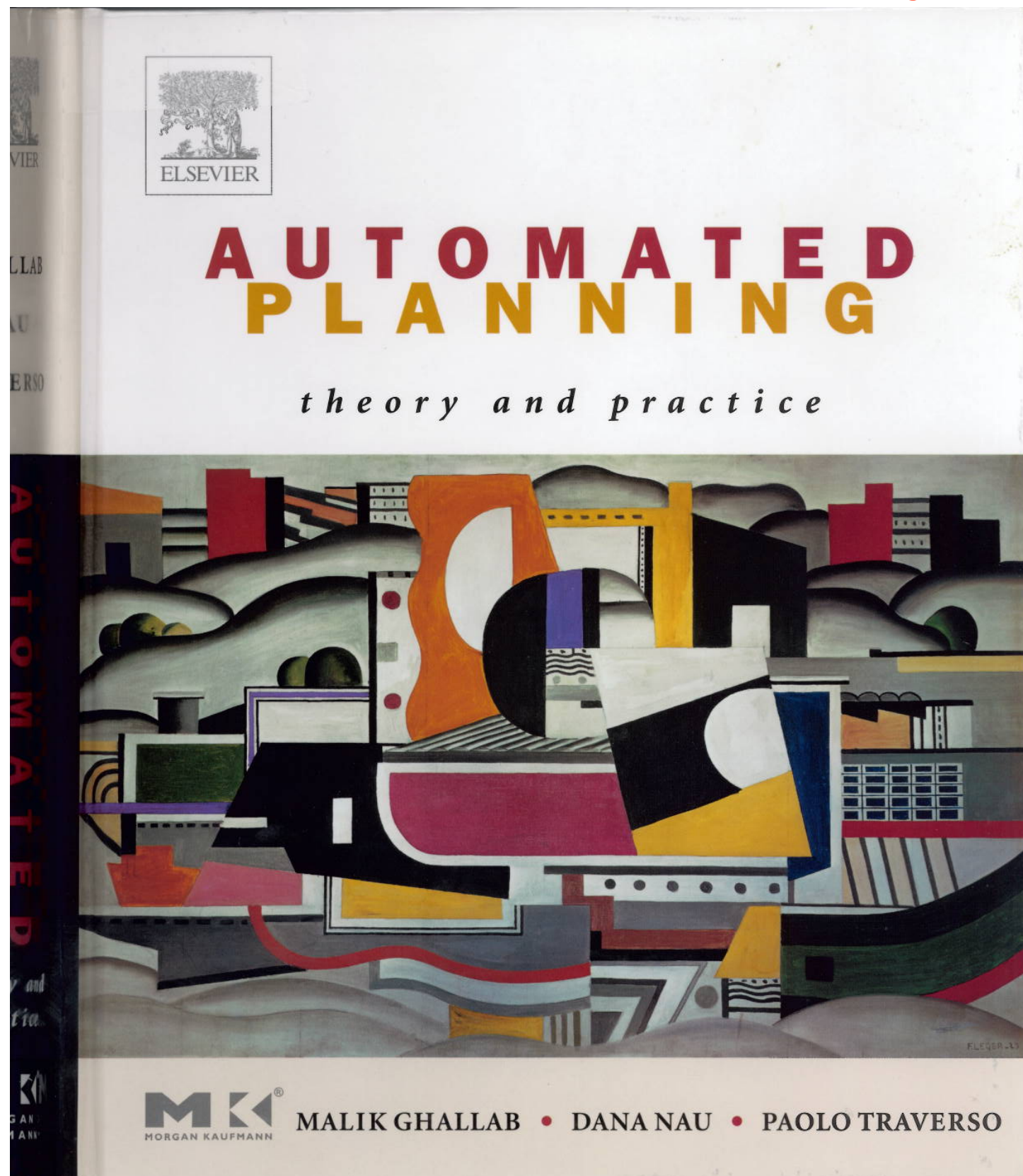
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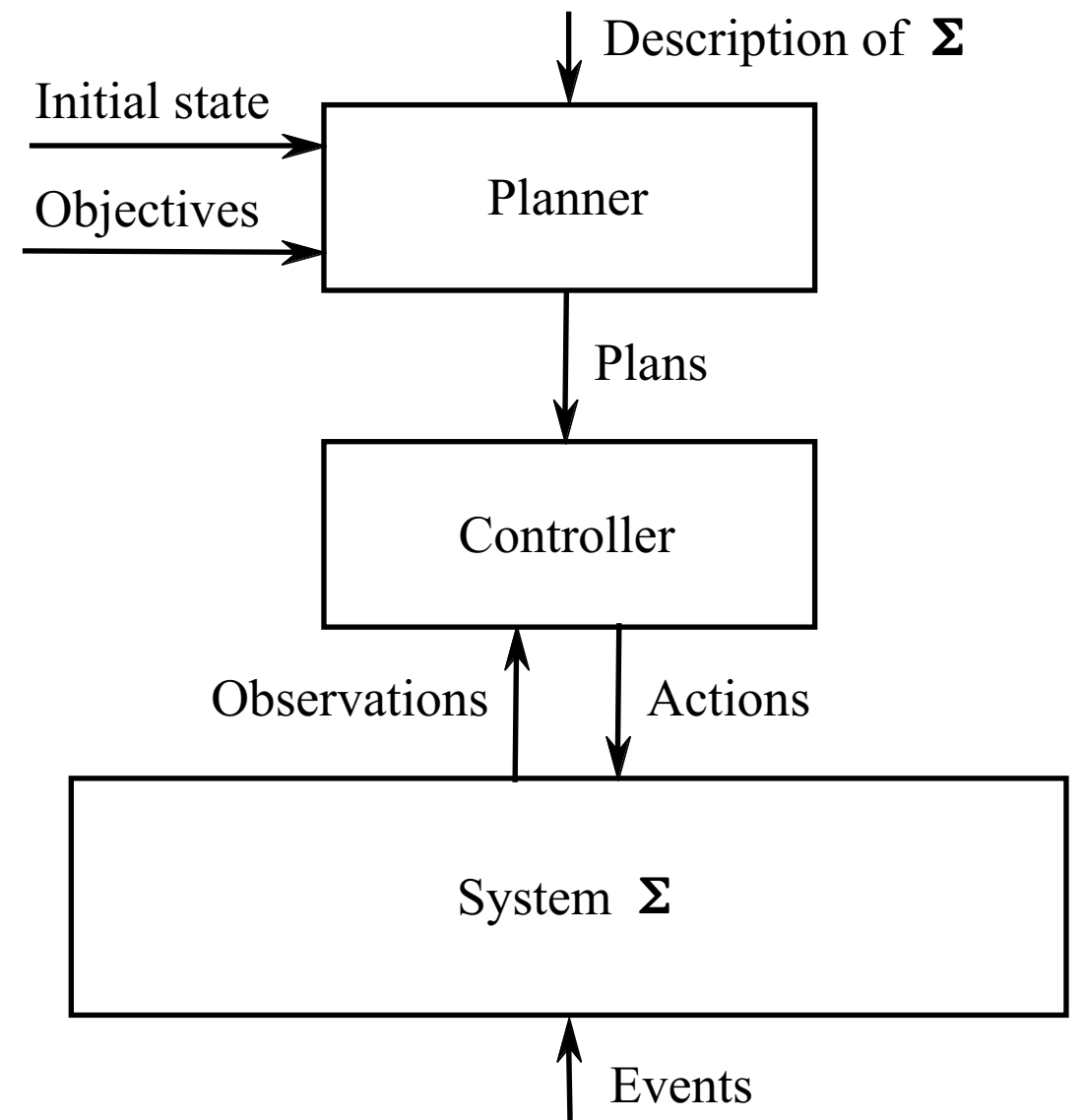
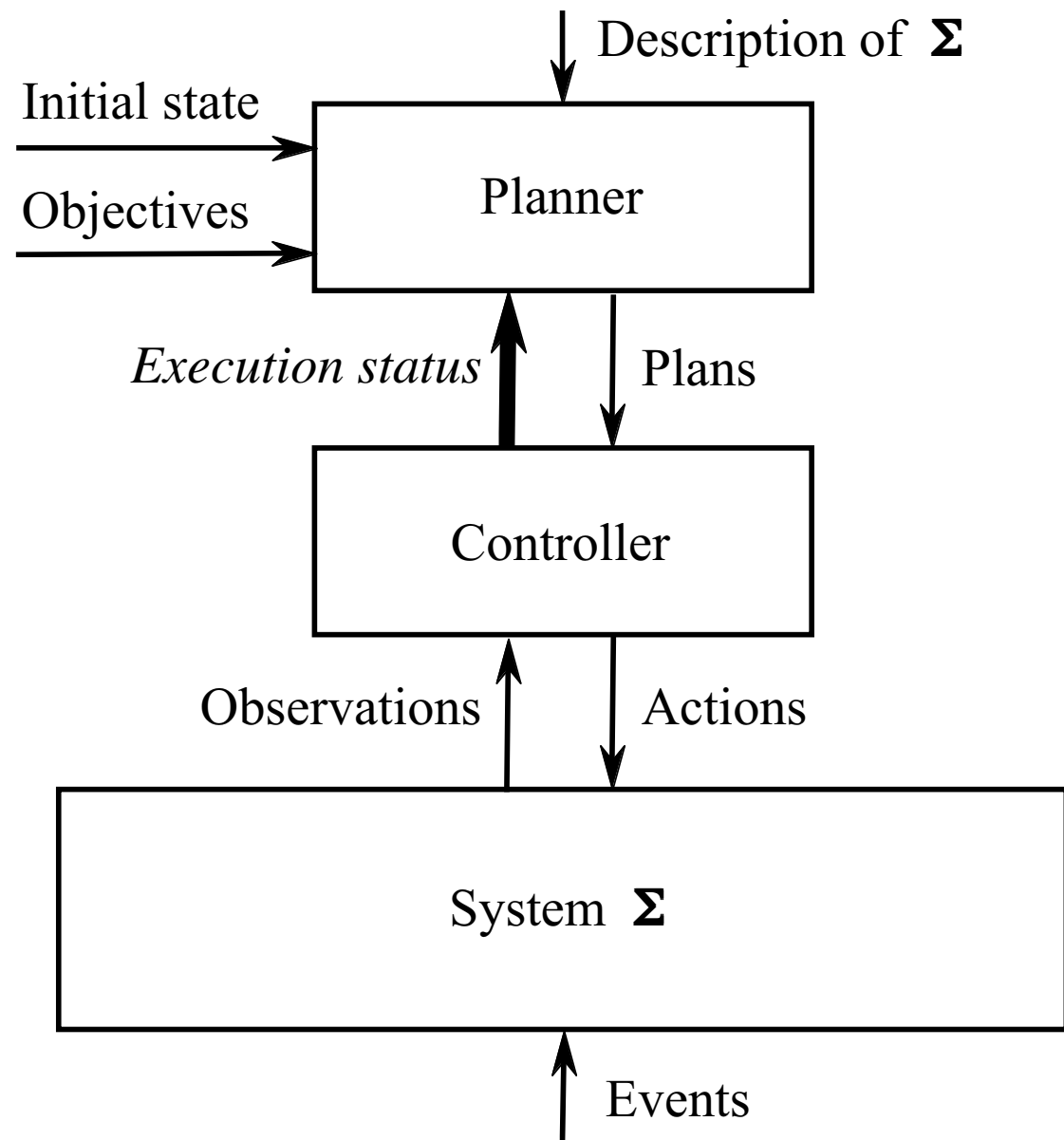


- ▶ Planing and Acting:
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 - 1.2 page / 630p.

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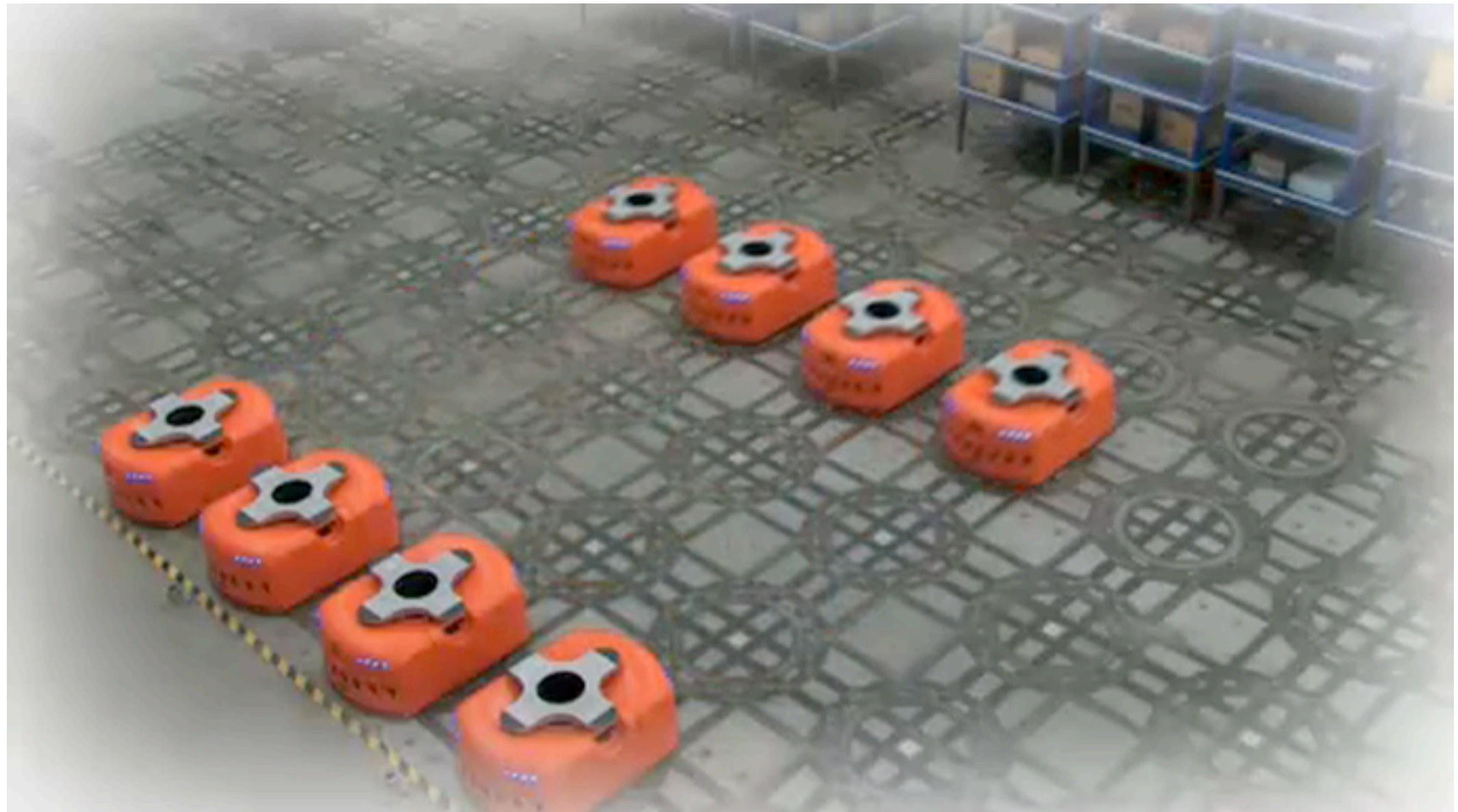
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- Robot manufacturers focus on
 - Programmable systems and ease of programming
 - Not much on autonomous deliberation through planning

Example of Warehouse automation



Kiva Systems

Household robot factotum

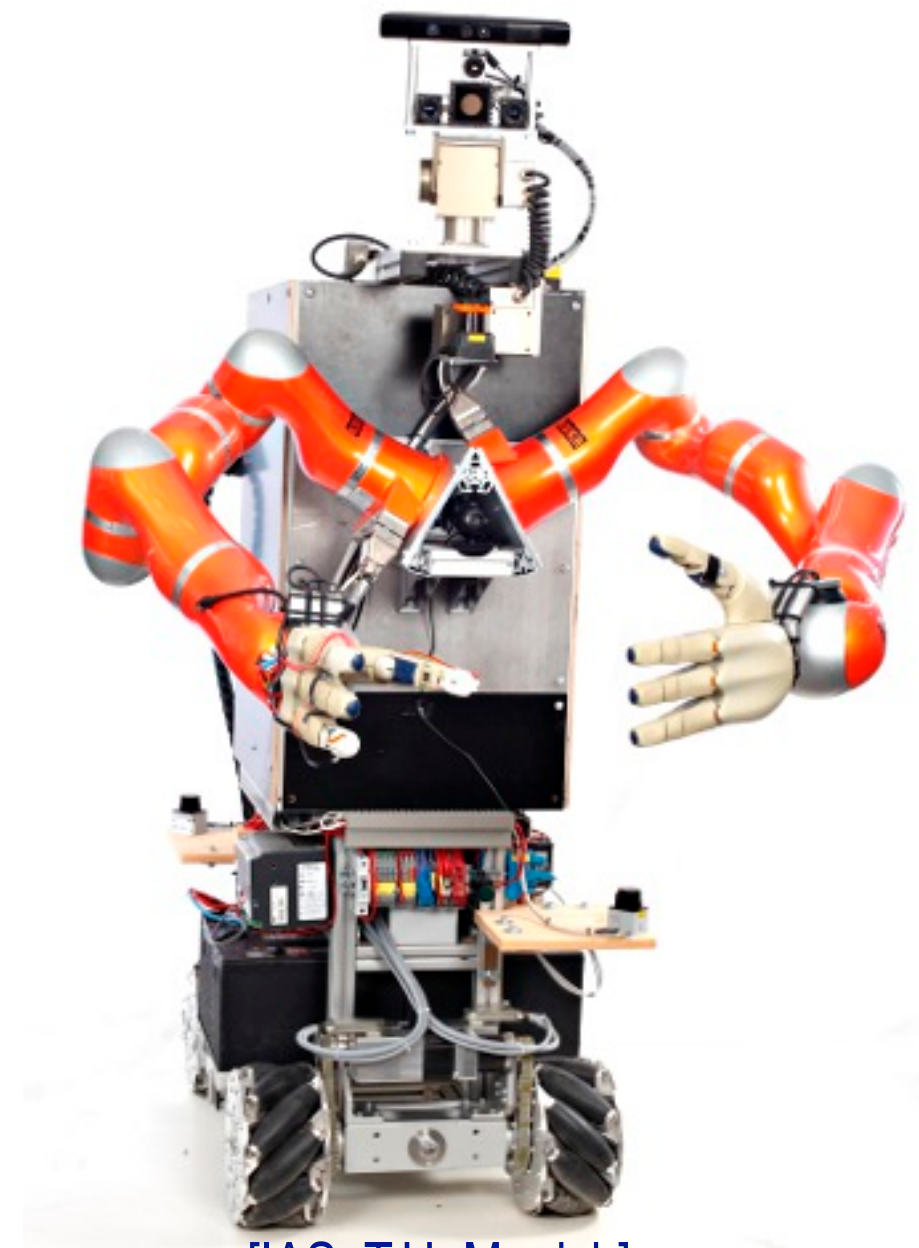


[LAAS, Toulouse]

Household robot factotum



[LAAS, Toulouse]



[IAS, T.U. Munich]

Household robot factotum

► Mission

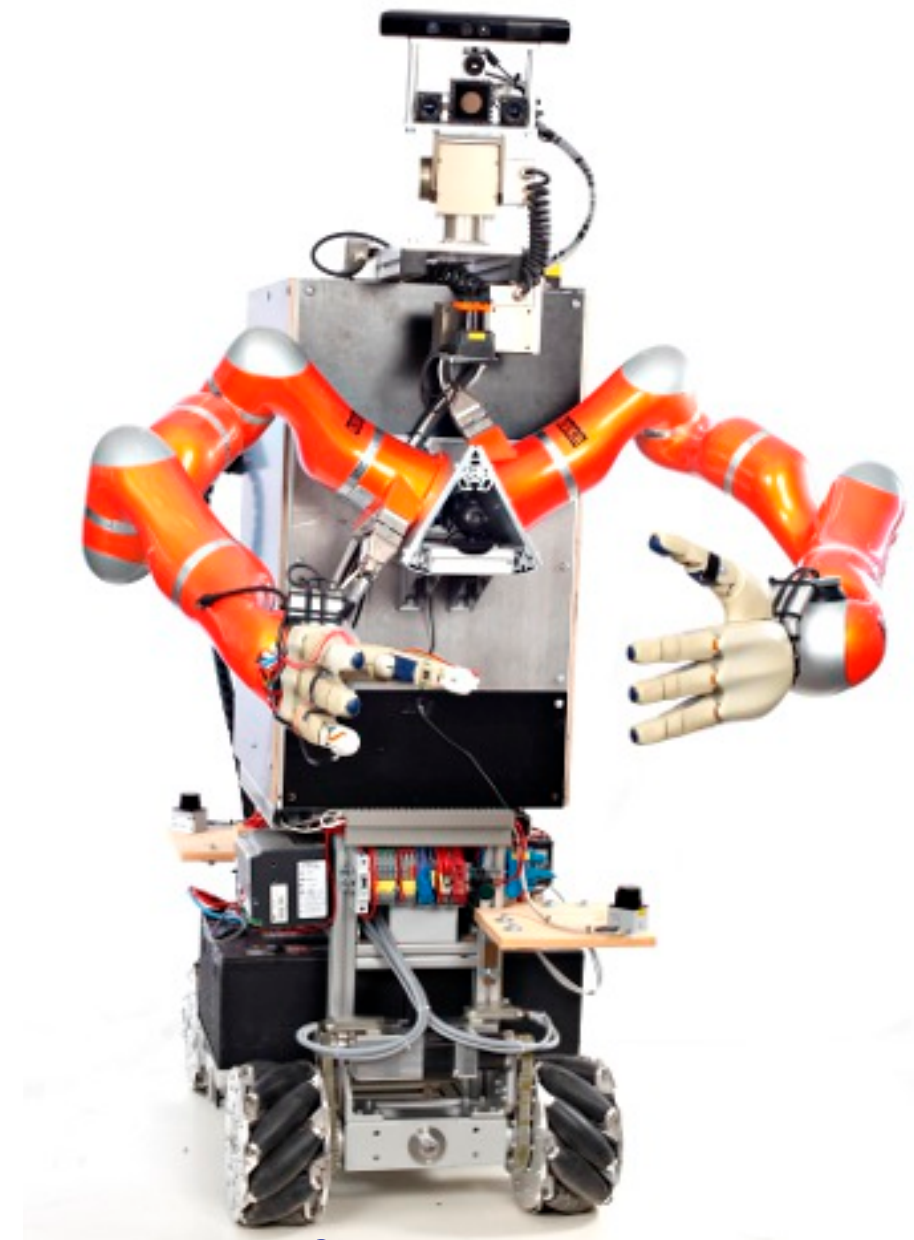
- Household services: pick, bring, maintain, sort, clean, cook
- Support users: help, remind, monitor

► Main features

- Wide *variety* of environments and tasks
- Complex interactions with users



[LAAS, Toulouse]



[IAS, T.U. Munich]

Basic skills for a factotum robot

- ▶ ***Localize*** itself in the environment map
- ▶ ***Map*** the environment: extend or update the map
- ▶ ***Move*** to a target position
- ▶ ***Detect*** and ***avoid*** obstacle
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Move: {environment features} → {navigation strategies}

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Mapping: dynamic & context-dependent

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(:action move

:parameters (?r - robot ?from ?to - location)

:precondition (and (adjacent ?from ?to)

(at ?r ?from) (not (occupied ?to)))

:effect (and (at ?r ?to) (not (occupied ?from))

(occupied ?to) (not (at ?r ?from))))

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$\text{move}(r, l, l')@ [t_s, t_e)$

precond: $\text{at}(r, l)@ [t_1, t_s), \text{free}(l')@ [t_2, t_e)$

effects: $\text{at}(r, \text{routes})@ [t_s, t_e)$

$\text{at}(r, l')@ [t_e, t_3)$

$\text{free}(l)@ [t_4, t_5)$

const: $t_s < t_4 < t_2, \text{adjacent}(l, l')$

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→ *the actor's view of deliberation*

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not enough addressed in planning

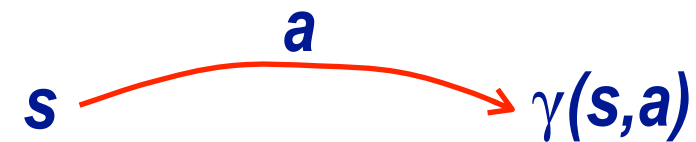
Why actor's view not enough addressed in planning

- ▶ Emphasis bias issue
 - Planning: Choice and organization of actions
 - Ingredients: Prediction + Search
 - *Precond-Effects* models
 - Focus on search
 - Trivialized prediction

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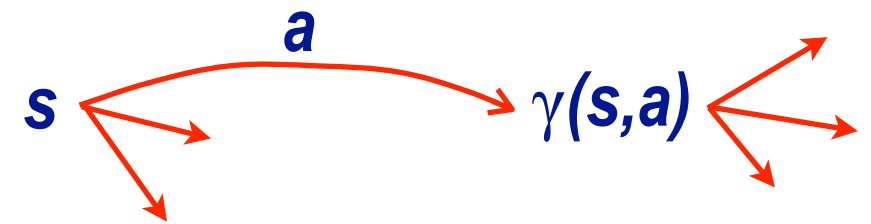
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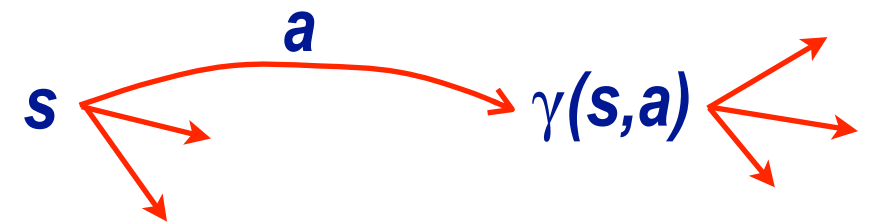
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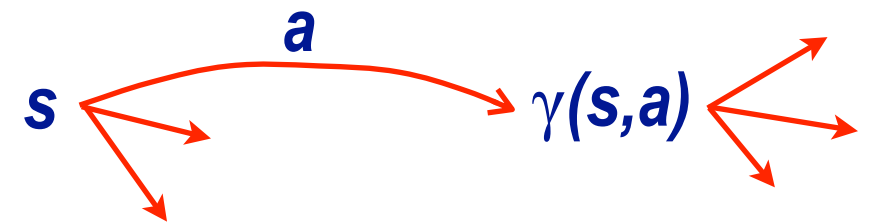
► Formalization issue

- Planning nicely formalized as mapping Domains \rightarrow Plans
- Acting difficult to formalize

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► Frontier issue

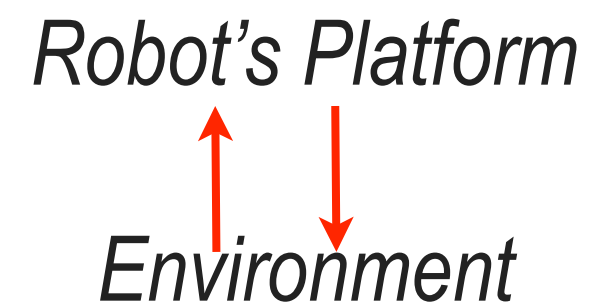
- Planning easily separated from sensory-motor functions
- “*Acting*” vs “*Executing*” more blurred frontier

Actor's view of deliberation

Actor's view of deliberation

Robot's Platform

Actor's view of deliberation



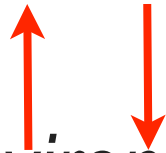
Actor's view of deliberation

Planning

Task \longrightarrow $\langle \dots, \text{abstract step}, \dots \rangle$

Robot's Platform

Environment




Actor's view of deliberation

Planning

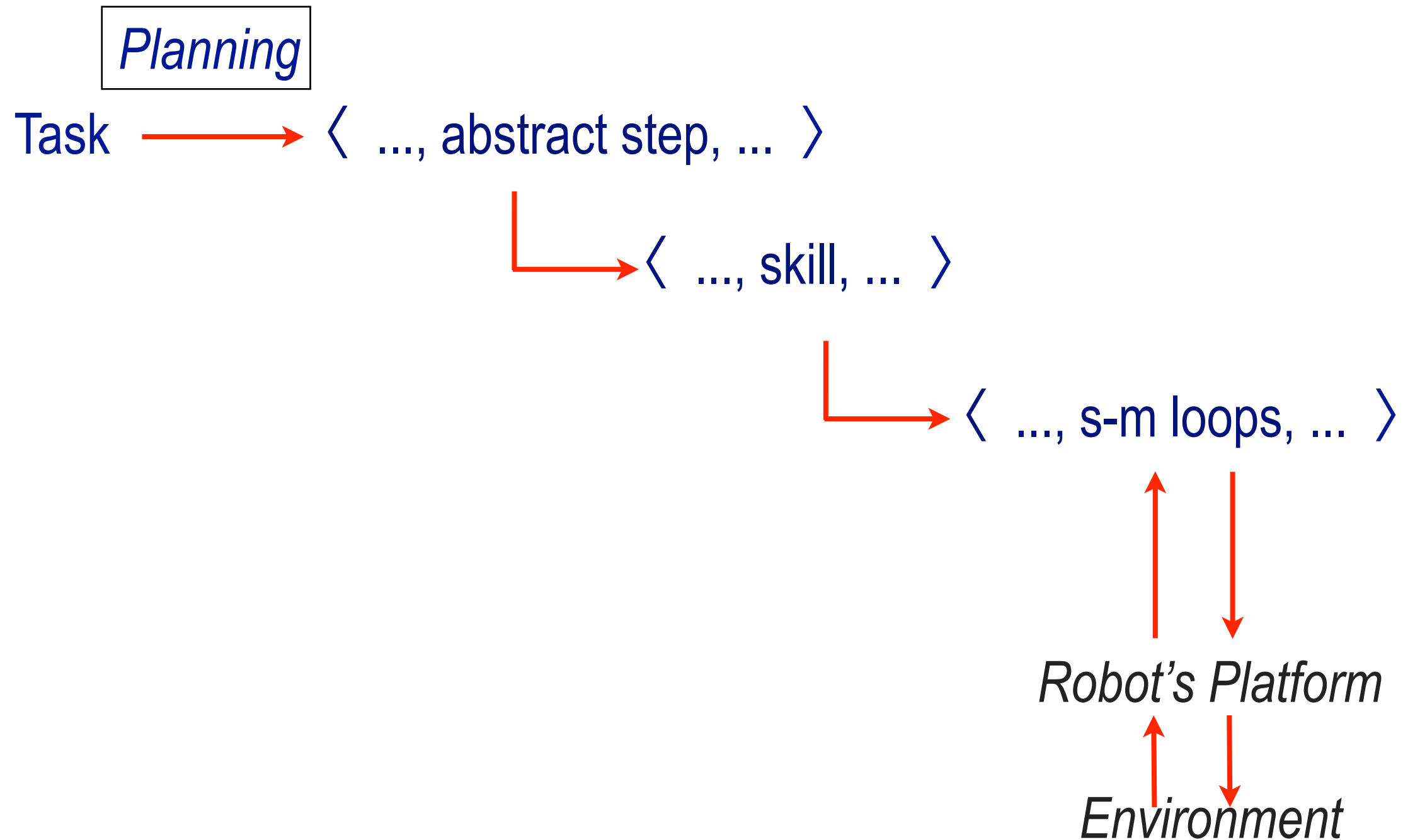
Task \longrightarrow $\langle \dots, \text{abstract step}, \dots \rangle$
 \searrow
 $\langle \dots, \text{skill}, \dots \rangle$

Robot's Platform

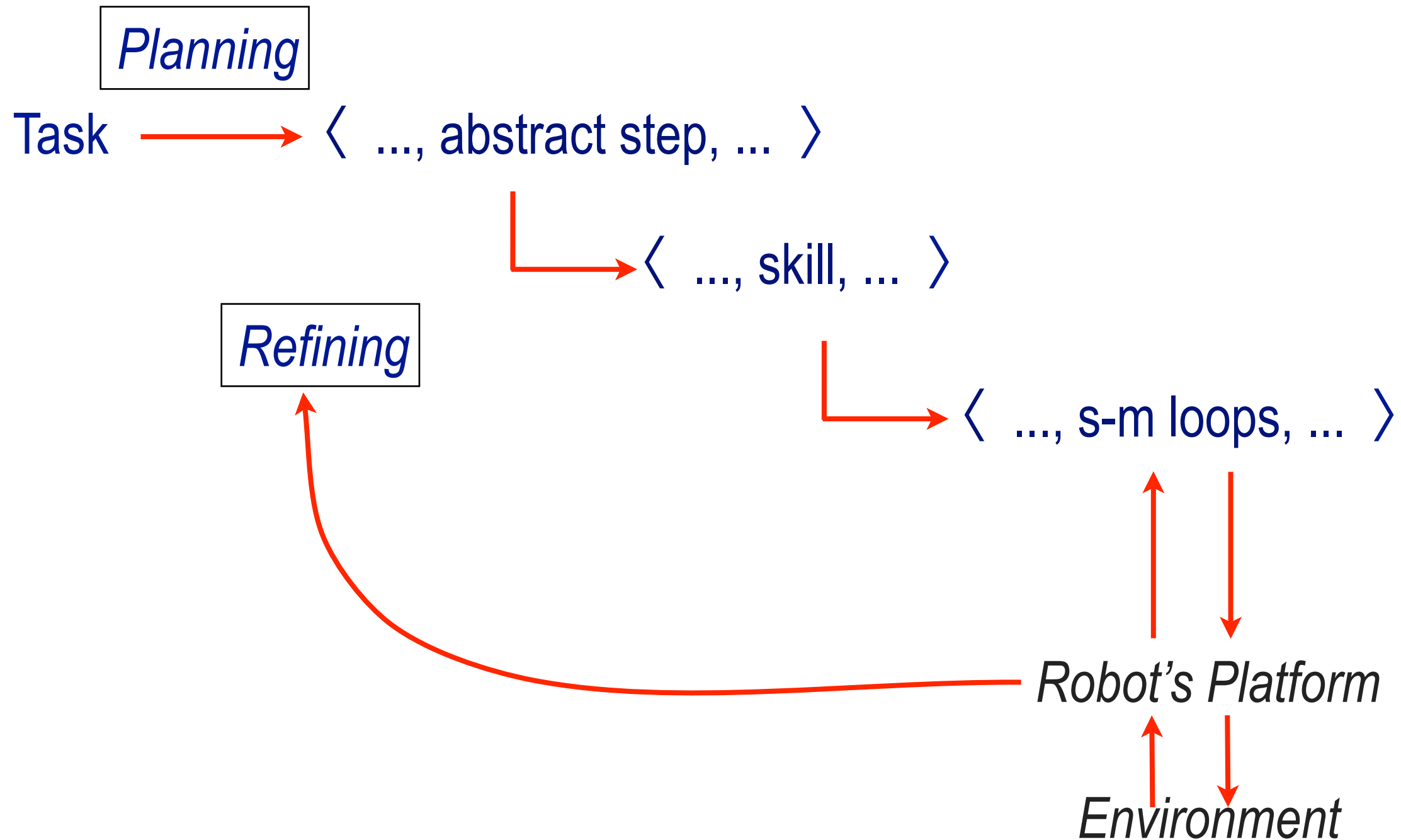


Environment

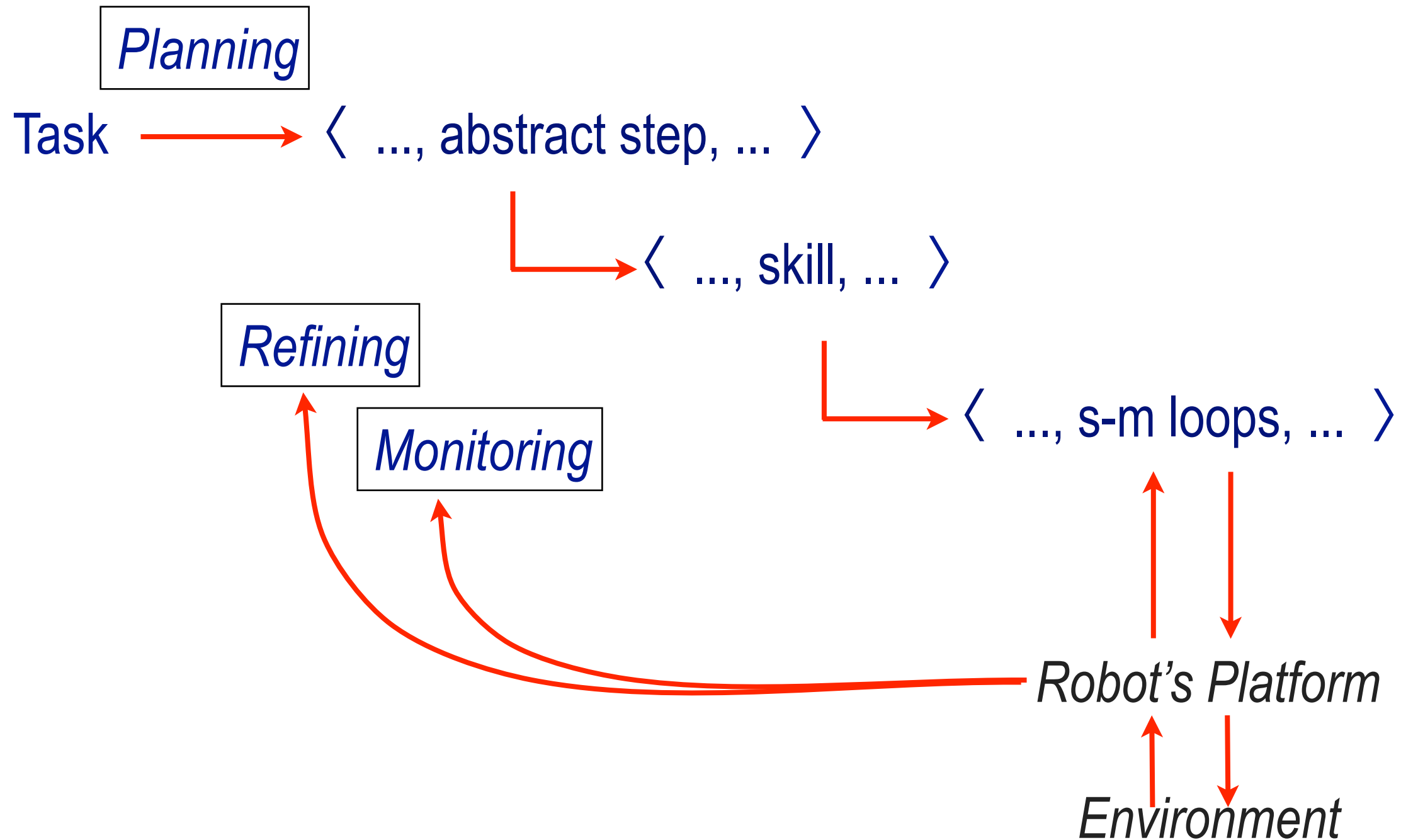
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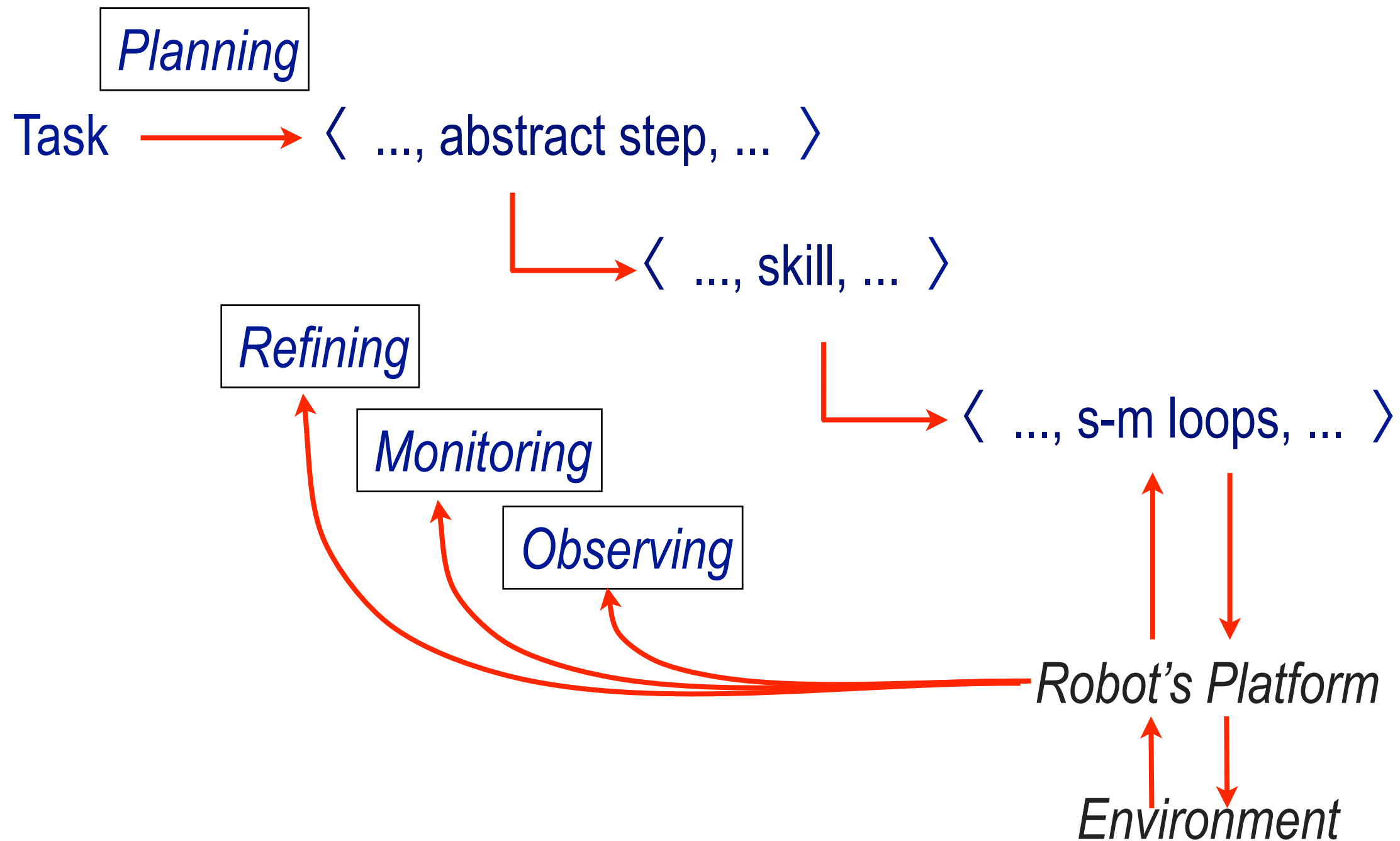
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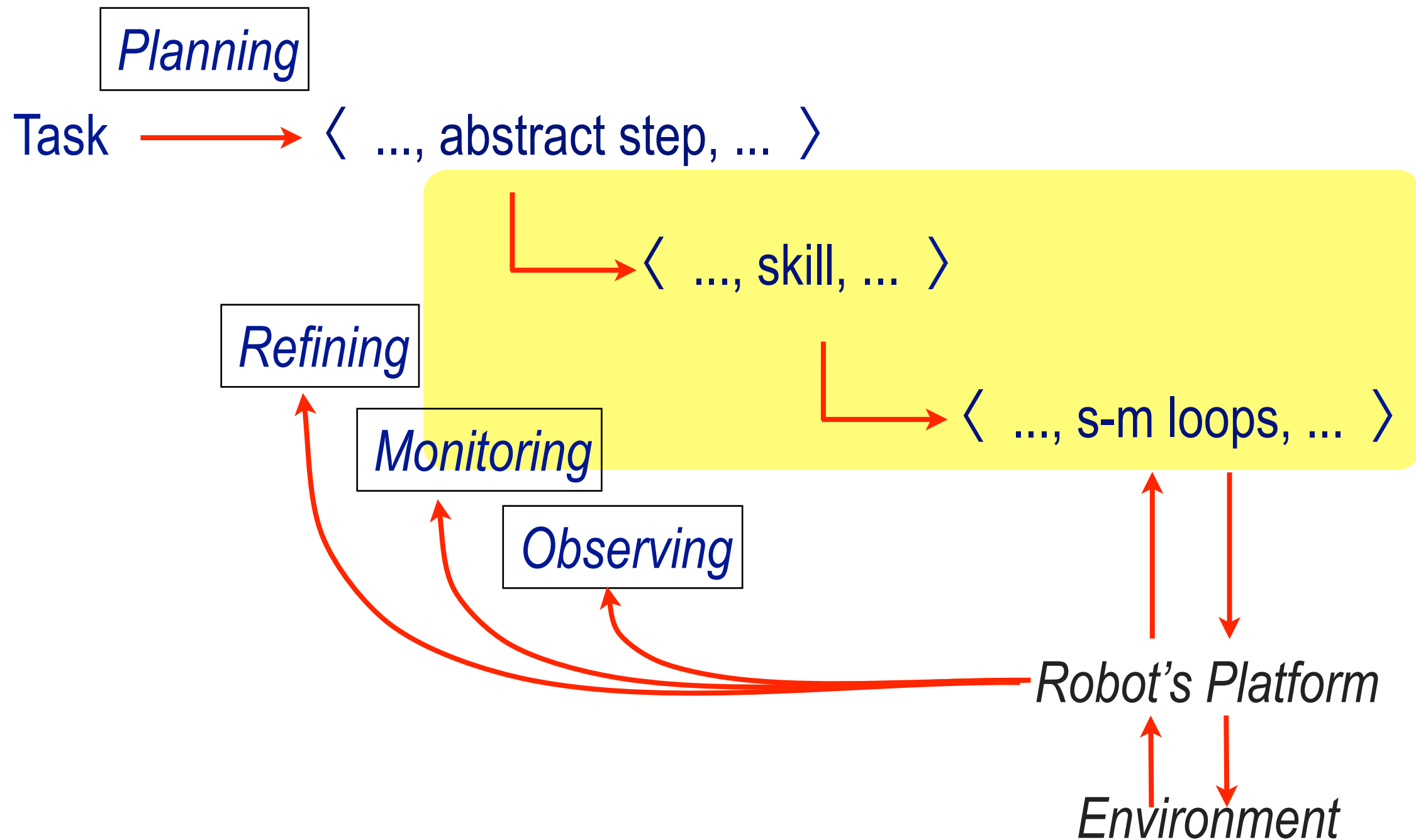
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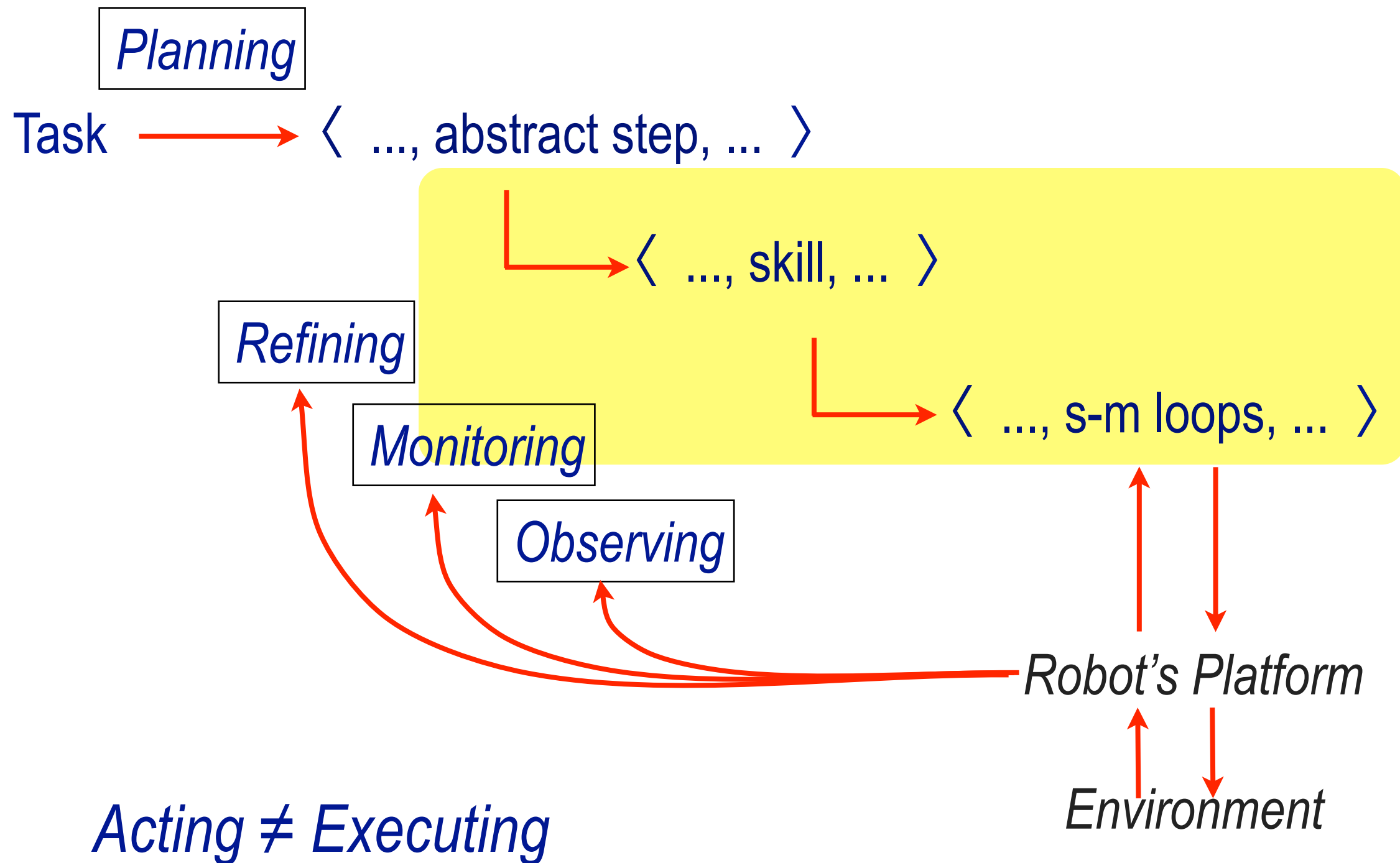
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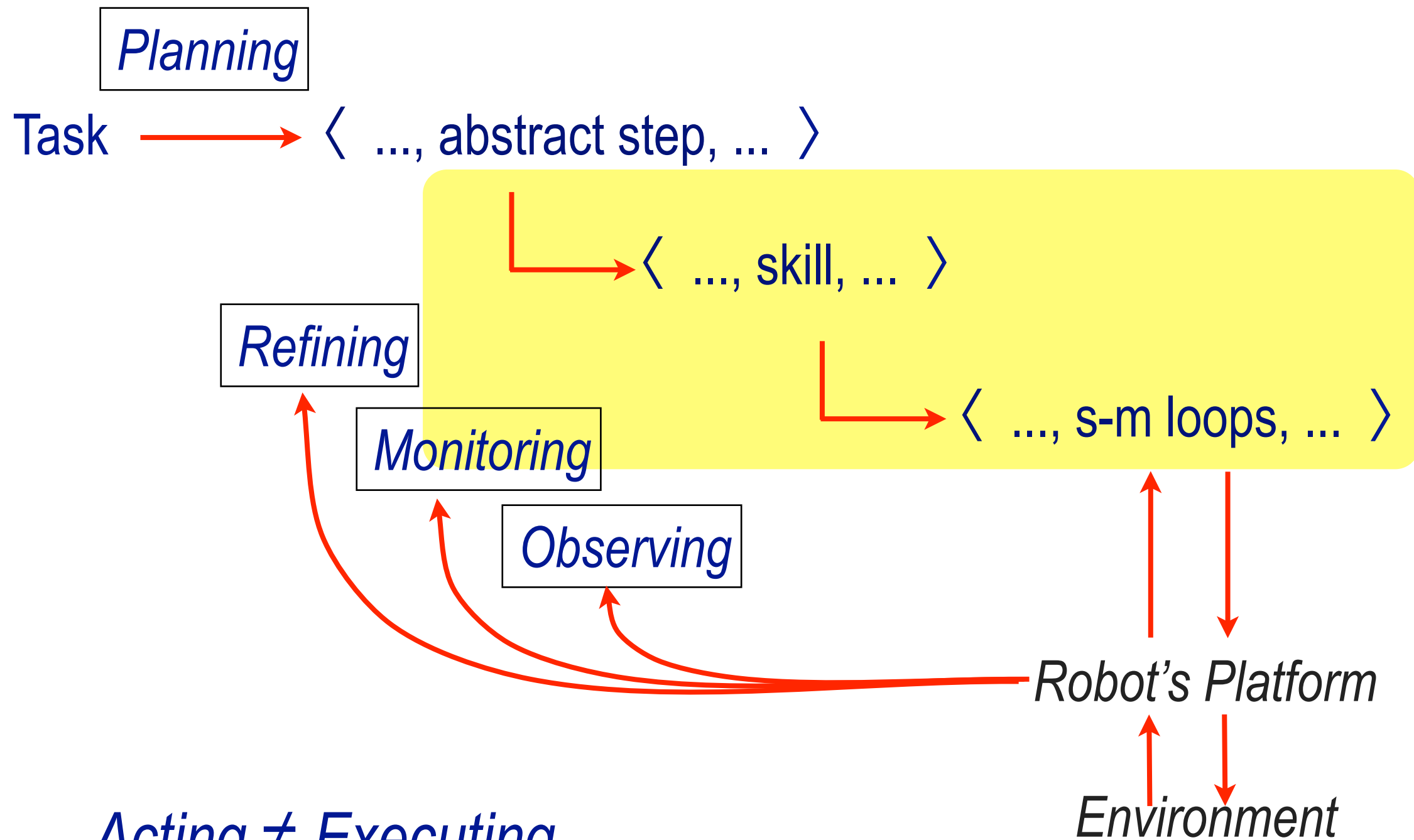
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Acting \neq Executing
Actor \neq Executor

✓ Motivations

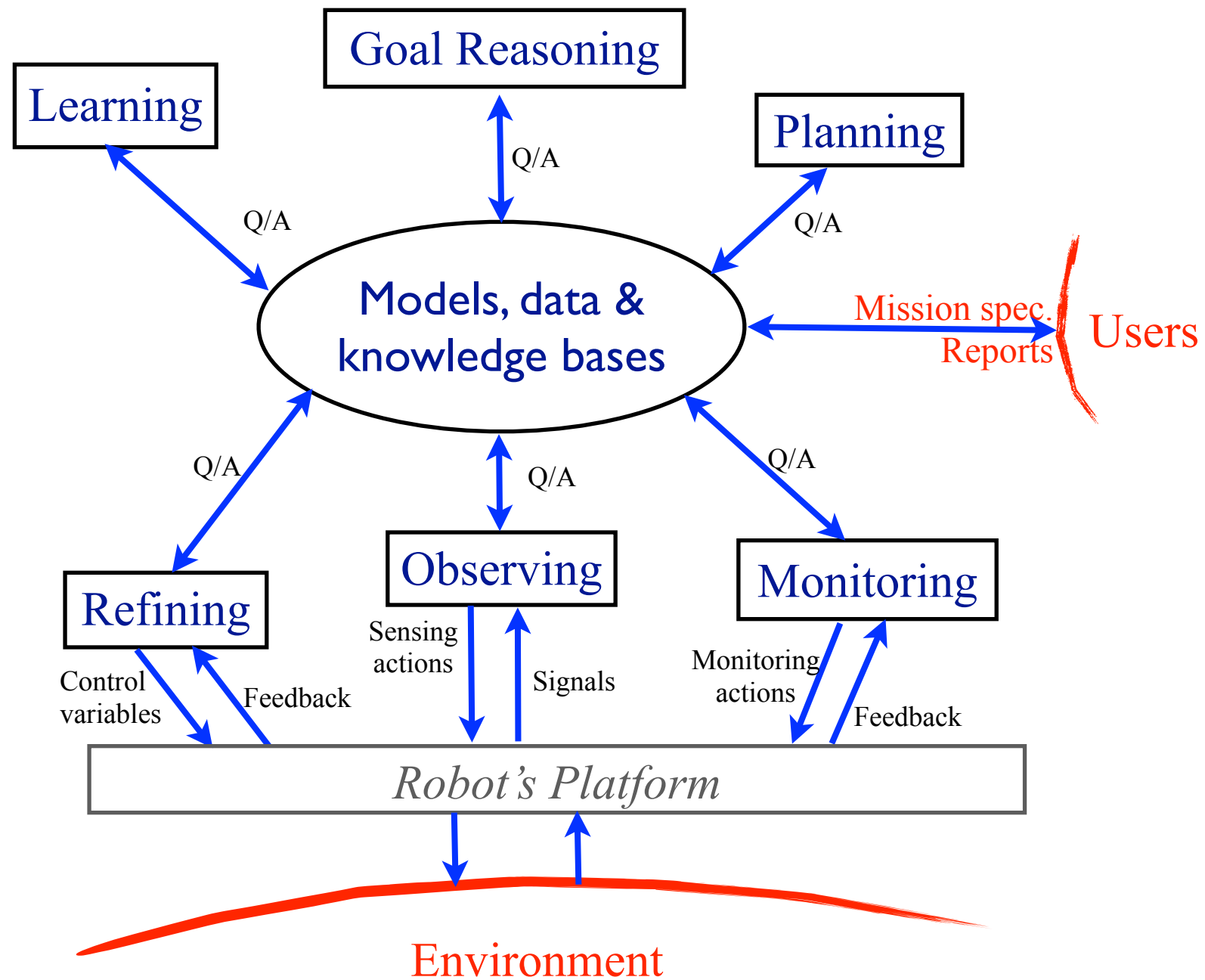
► Deliberation functions

- Planning
- Refining
- Monitoring
- Perceiving
- Goal reasoning
- Learning
- Integration

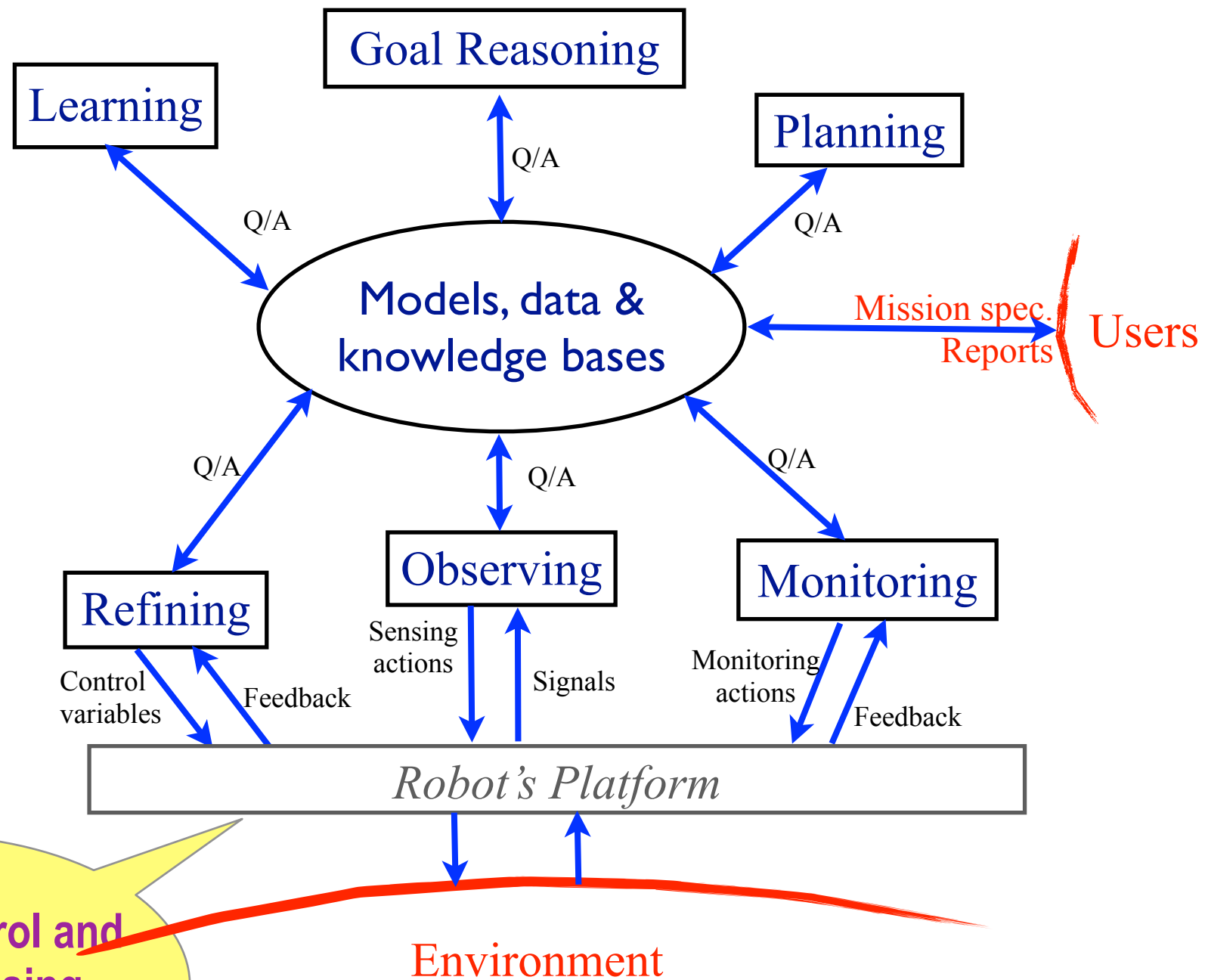
► Research Challenges

- Representation
- Model acquisition & Verification
- Synthesis
- Monitoring and Goal reasoning
- Integration

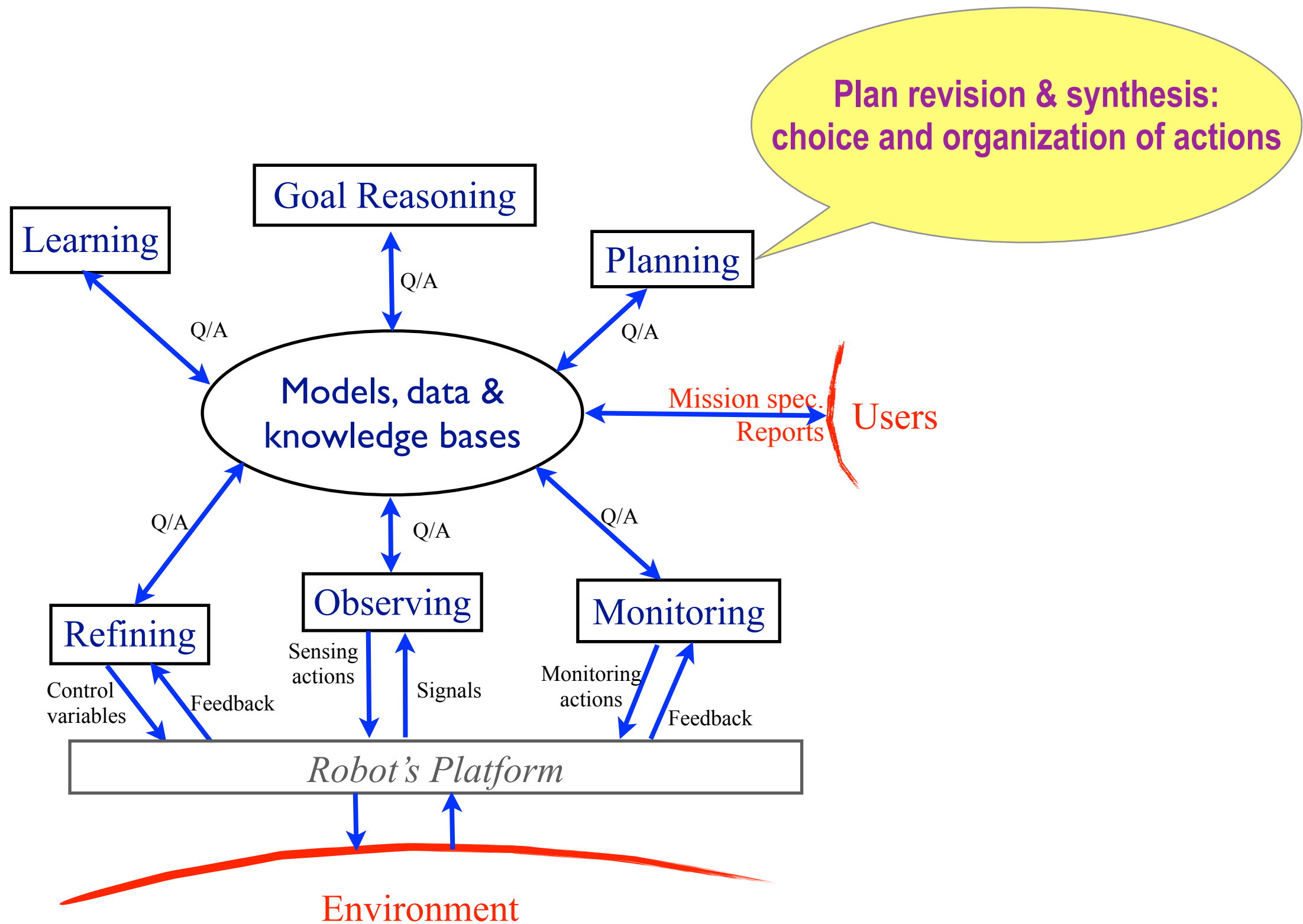
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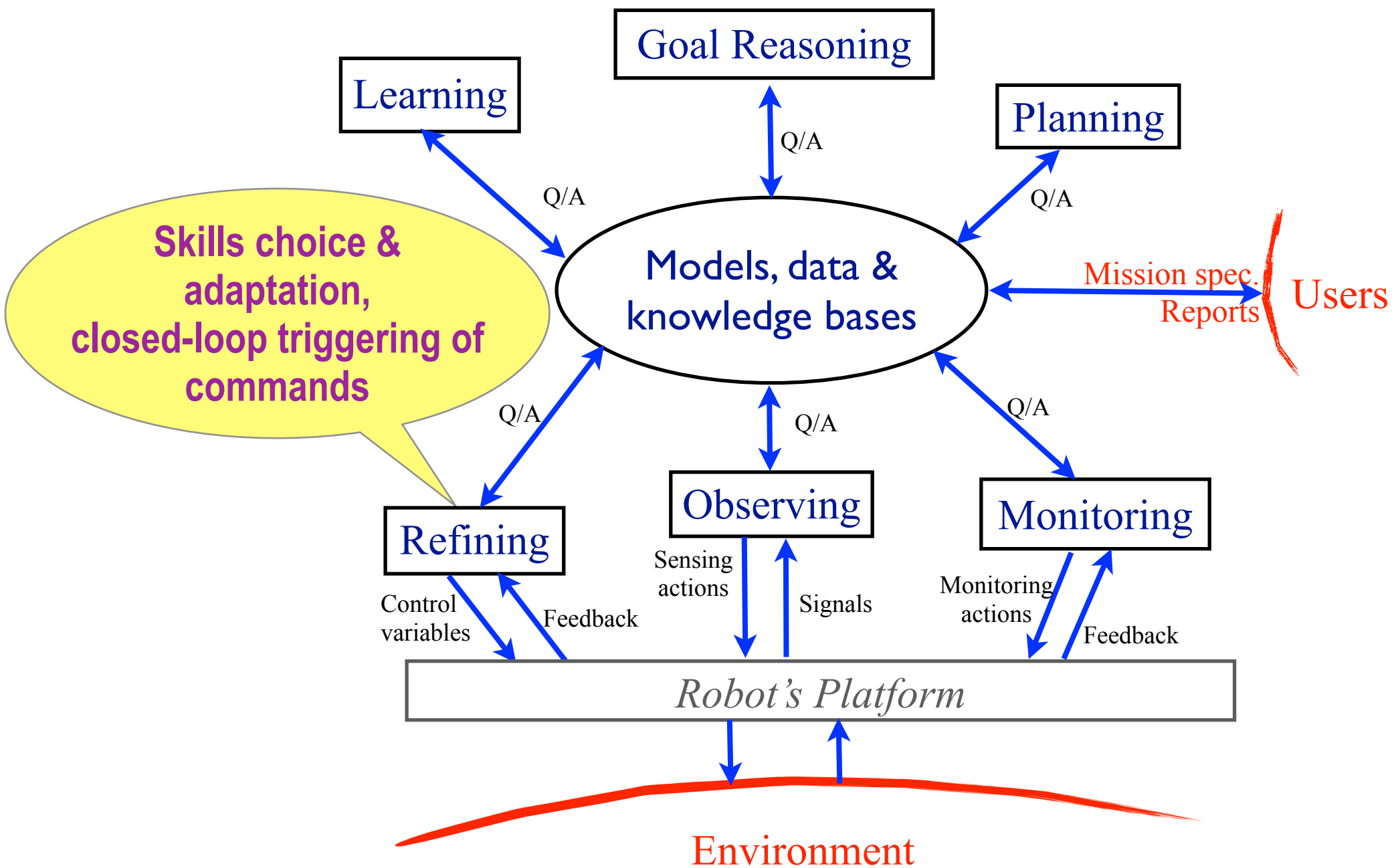
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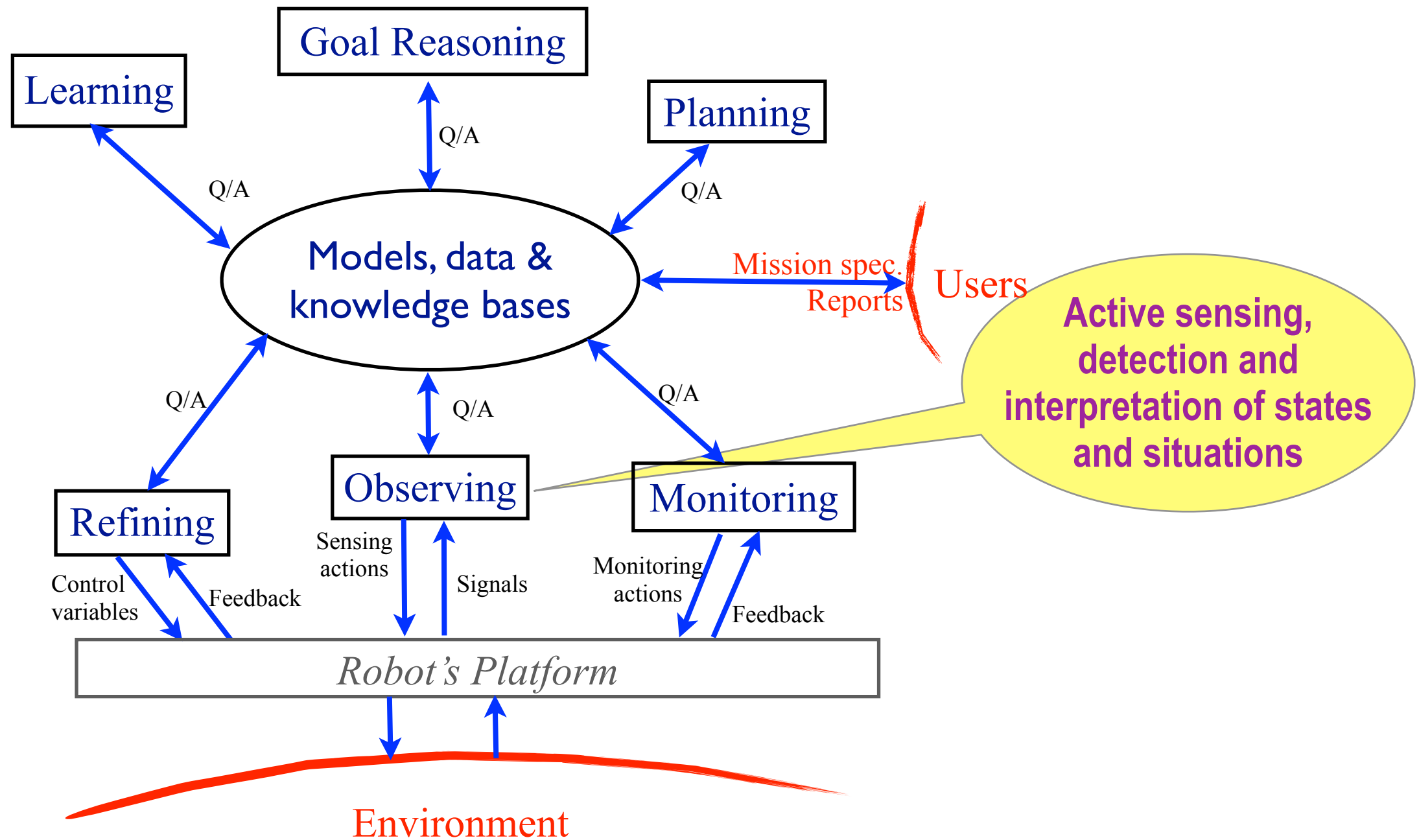
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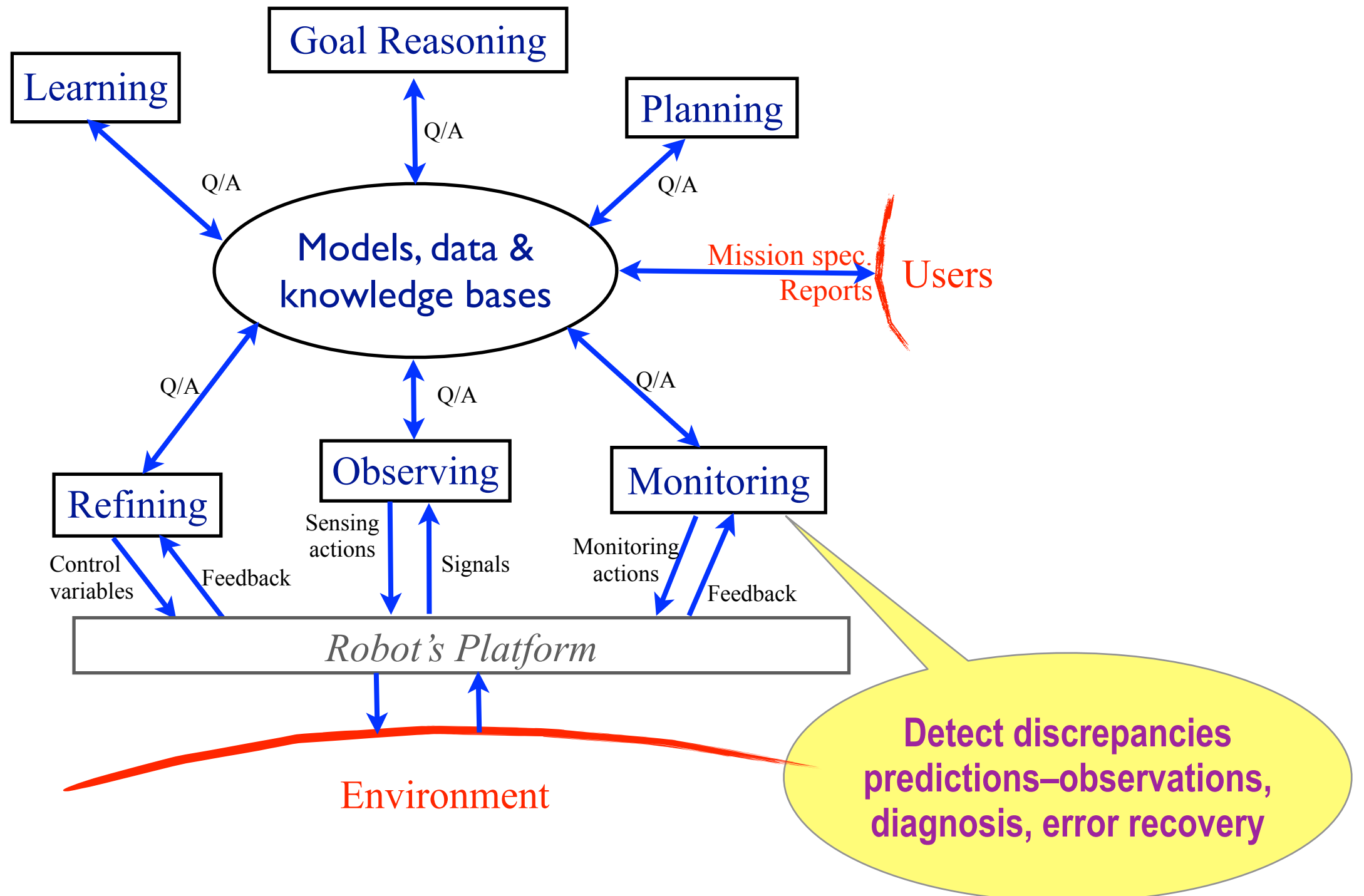
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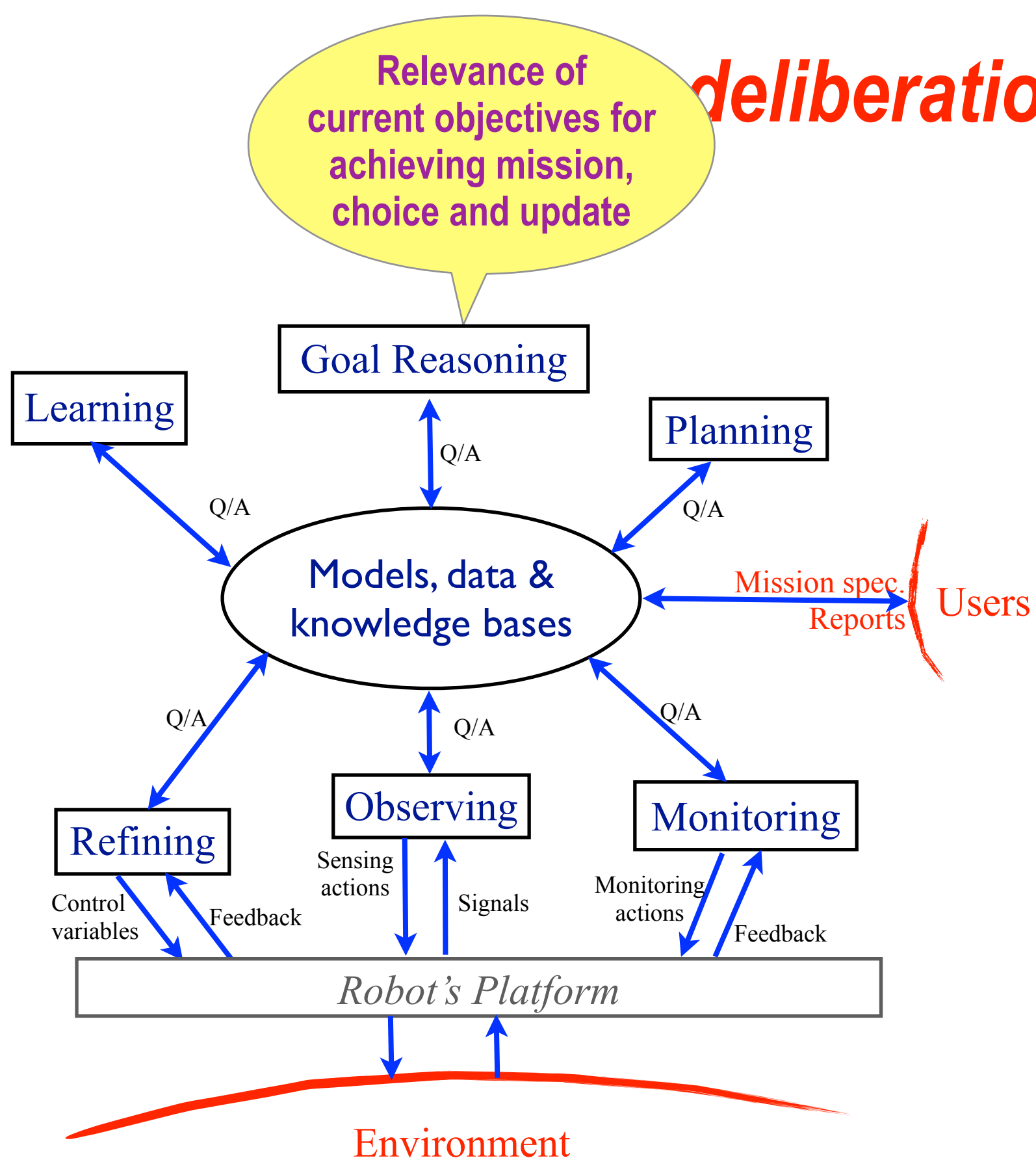
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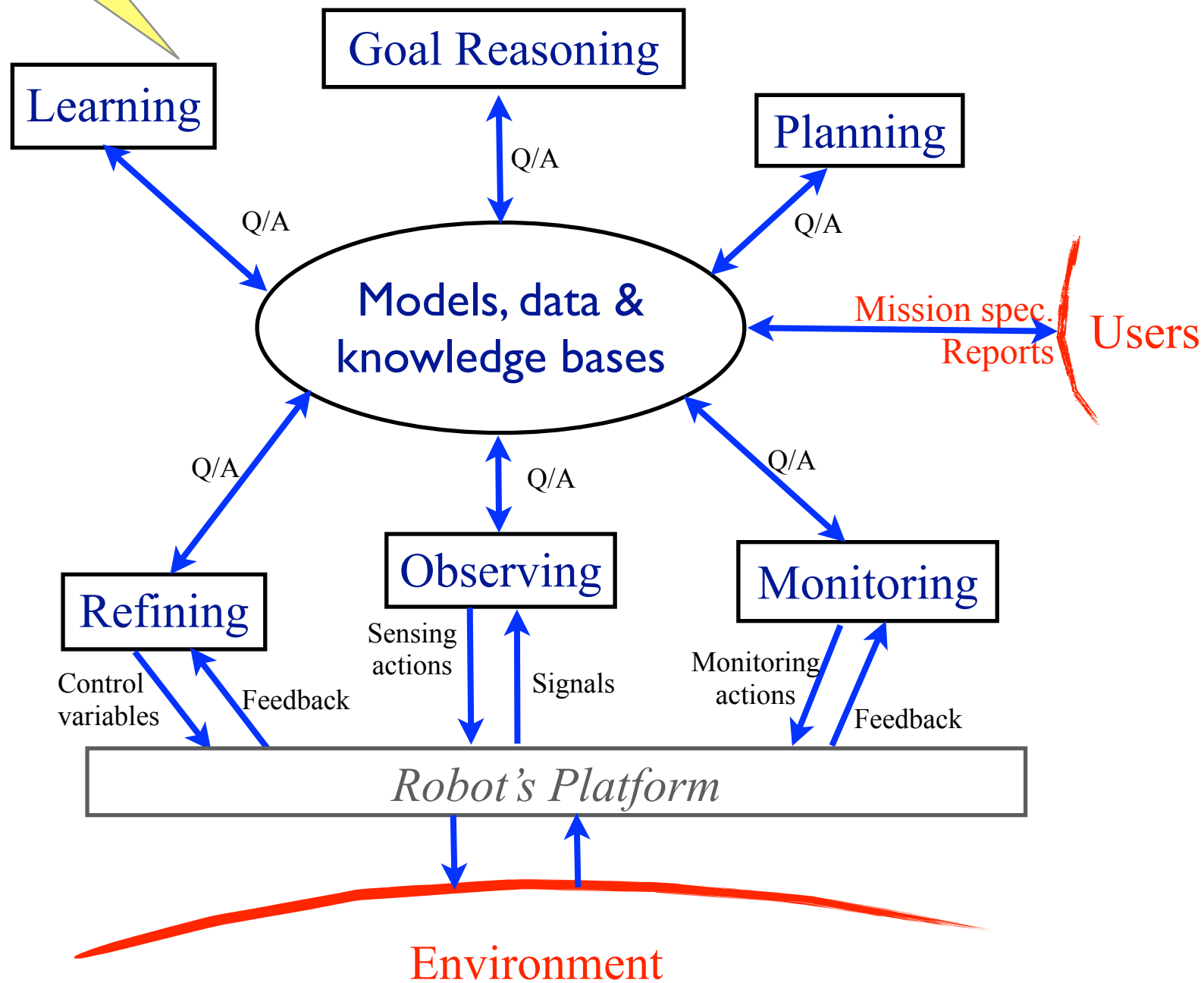


deliberation functions

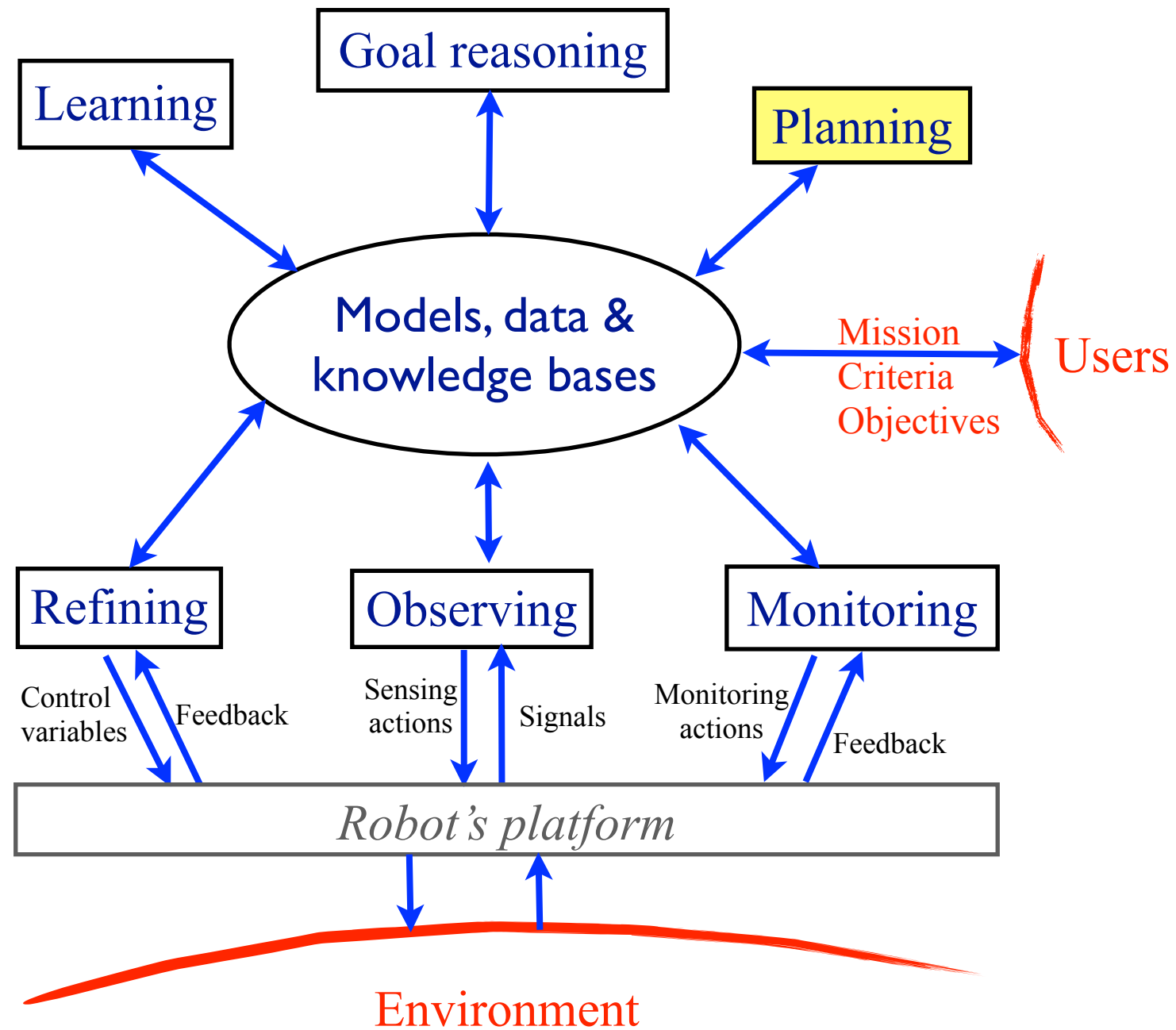


Actor's deliberation functions

Active learning to
acquire and improve
models and behaviors



Actor's deliberation functions



► *Different types of actions*

=> Different predictive models

=> Different planning problems and techniques

- Motion and manipulation planning
- Perception planning
- Navigation planning
- Communication planning
- Task planning

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➡ Integration of planners with distinct state and action spaces

Planning in Robotics

▶ *Usual context*

- Open, dynamic environment
- Interaction with users
- Online, interleaved with acting

▶ *Desirable features*

- Access to external domain knowledge
- Manage concurrency
- Reason on uncertainty

▶ *Approaches*

- Hierarchical
- Temporal
- Probabilistic

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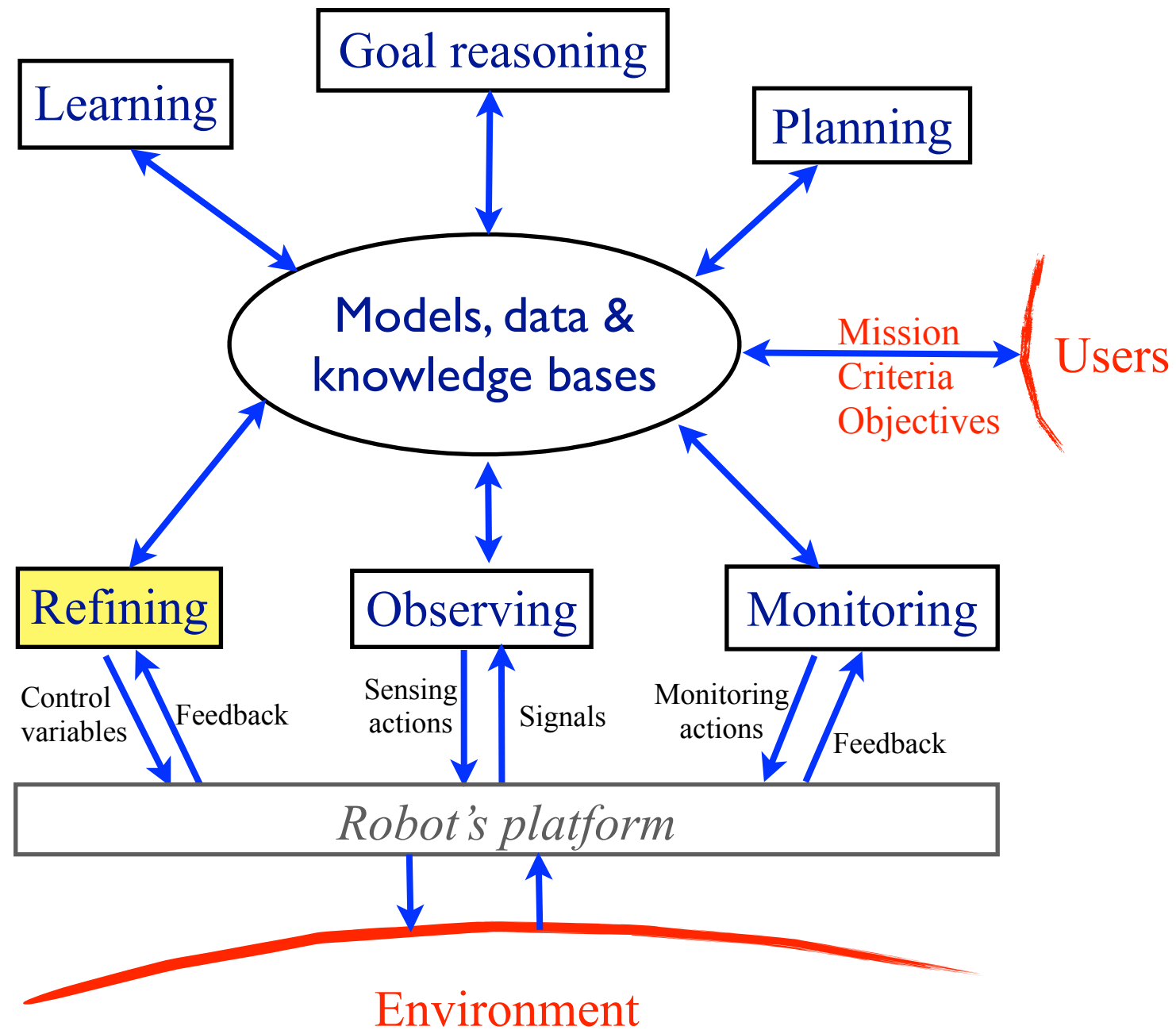
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► *Approaches*

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- Probabilistic
- => Sparse probabilistic models

Actor's deliberation functions



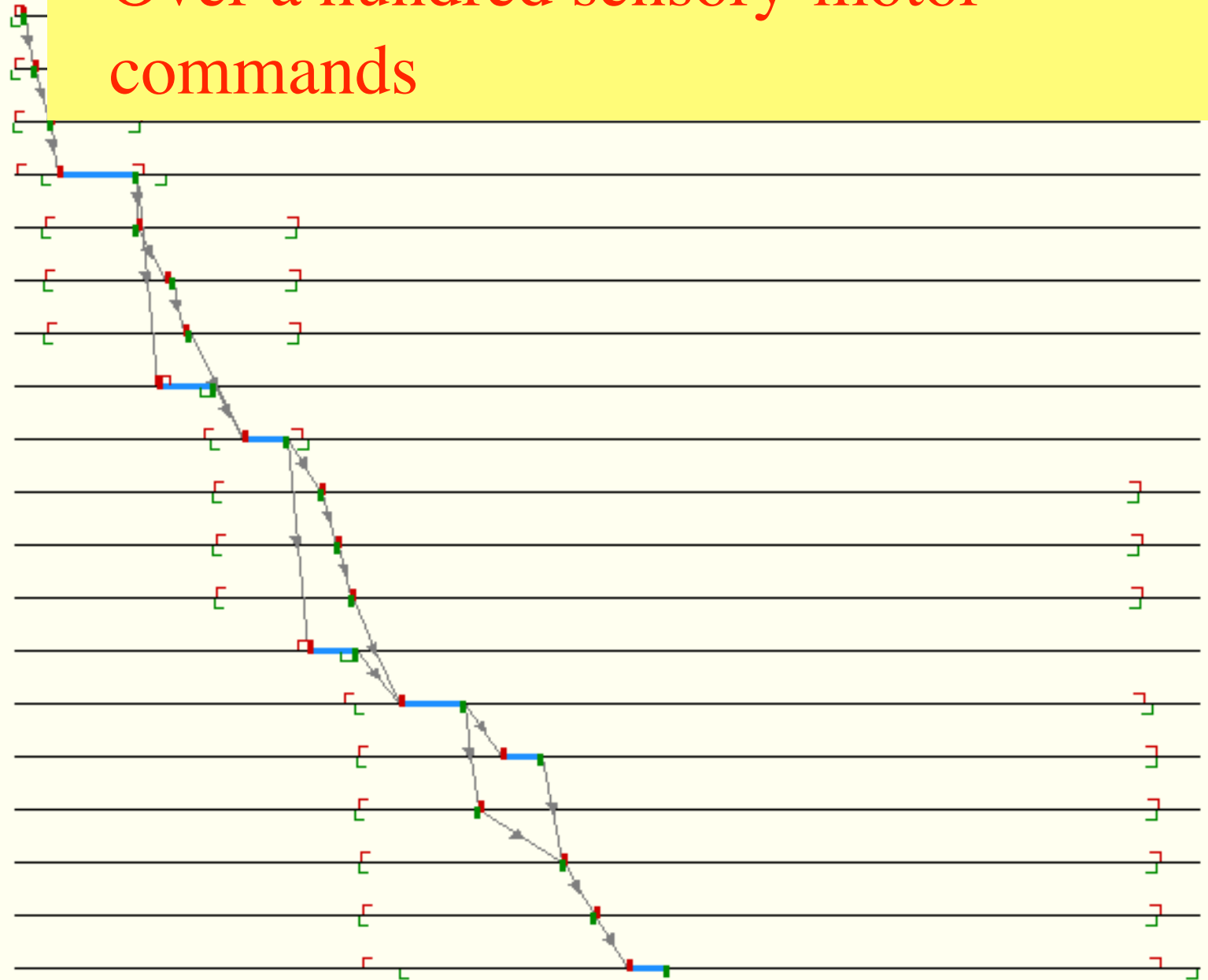
Refining abstract steps into skills

- About 20 plan steps
- Over a hundred sensory-motor commands

X Tasks Table: plan.ix

Mouse Time:23:59

```
MOVE_PAN_TILT_UNIT(STRAIGHT, AT_MY_FEET)
TAKE_PICTURE(OBJ1, 0.500000, -0.500000)
MOVE_PAN_TILT_UNIT(AT_MY_FEET, STRAIGHT)
MOVEX(0.500000, -0.500000, 9.000000, -0.500000)
MOVE_PAN_TILT_UNIT(STRAIGHT, AT_MY_FEET)
TAKE_PICTURE(OBJ2, 9.000000, -0.500000)
MOVE_PAN_TILT_UNIT(AT_MY_FEET, STRAIGHT)
COMMUNICATE(W1)
MOVE(9.000000, -0.500000, 10.000000, -3.000000)
MOVE_PAN_TILT_UNIT(STRAIGHT, AT_MY_FEET)
TAKE_PICTURE(OBJ4, 10.000000, -3.000000)
MOVE_PAN_TILT_UNIT(AT_MY_FEET, STRAIGHT)
COMMUNICATE(W2)
MOVE(10.000000, -3.000000, 8.000000, -5.000000)
DOWNLOAD_IMAGES()
MOVE_PAN_TILT_UNIT(STRAIGHT, AT_MY_FEET)
TAKE_PICTURE(OBJ3, 8.000000, -5.000000)
MOVE_PAN_TILT_UNIT(AT_MY_FEET, STRAIGHT)
MOVE(8.000000, -5.000000, 0.500000, -0.500000)
```



Refining abstract steps into skills

20 abstract steps:
Descriptive models
What, When, Why

Refinement



100 s-m commands:
Operational models
Modus operandi
How

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How to get the operational models ?

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How to get the operational models ?

- Specification

Refining abstract steps into skills

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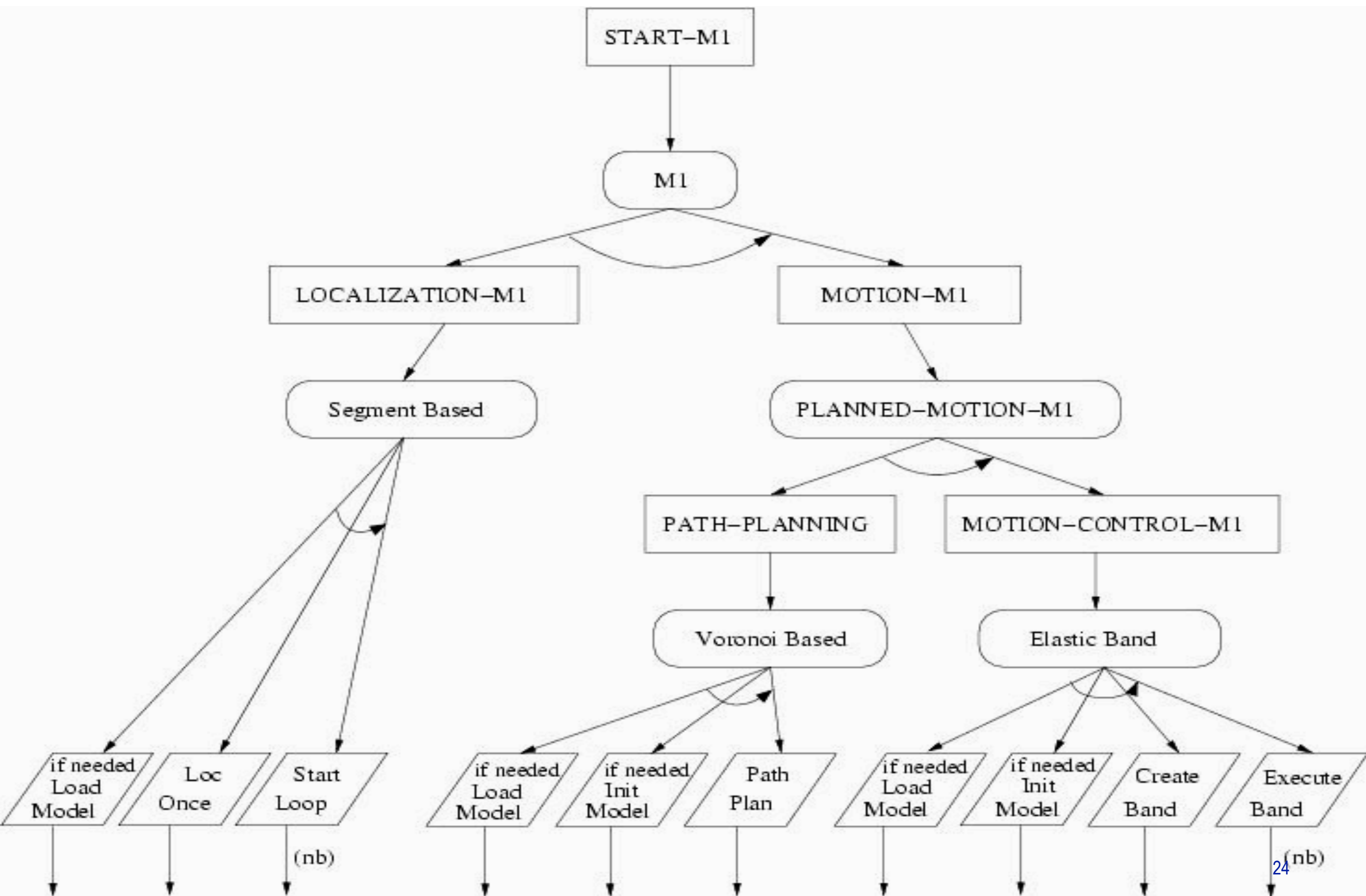


100 s-m commands:
Operational models
Modus operandi
How

How to get the operational models ?

- Specification
- Synthesis

Synthesis of skills achieving an action



Approaches

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- ▶ Informal specification
PRS, RAP, TDL, XFRM
 - Flexible
 - Hand-written

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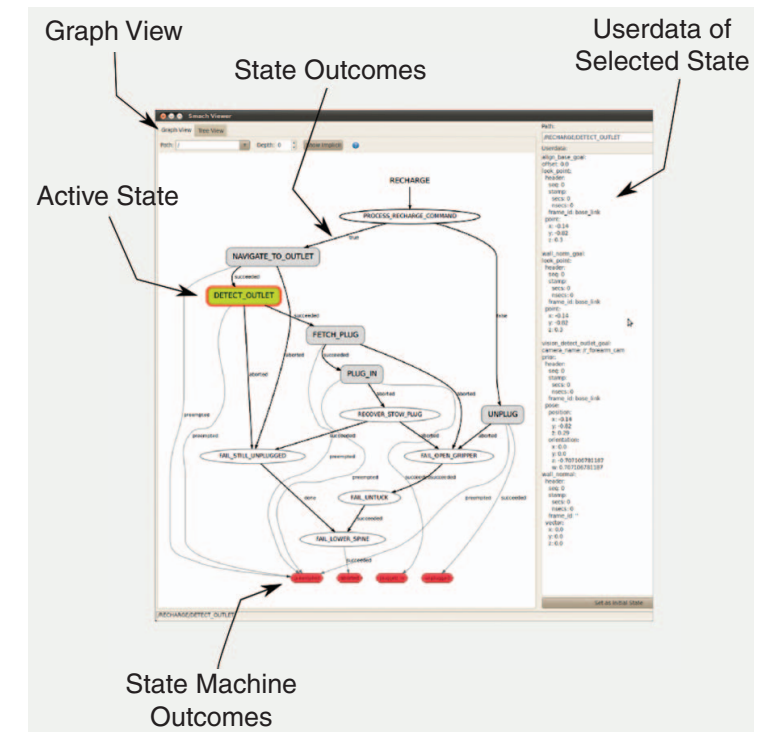
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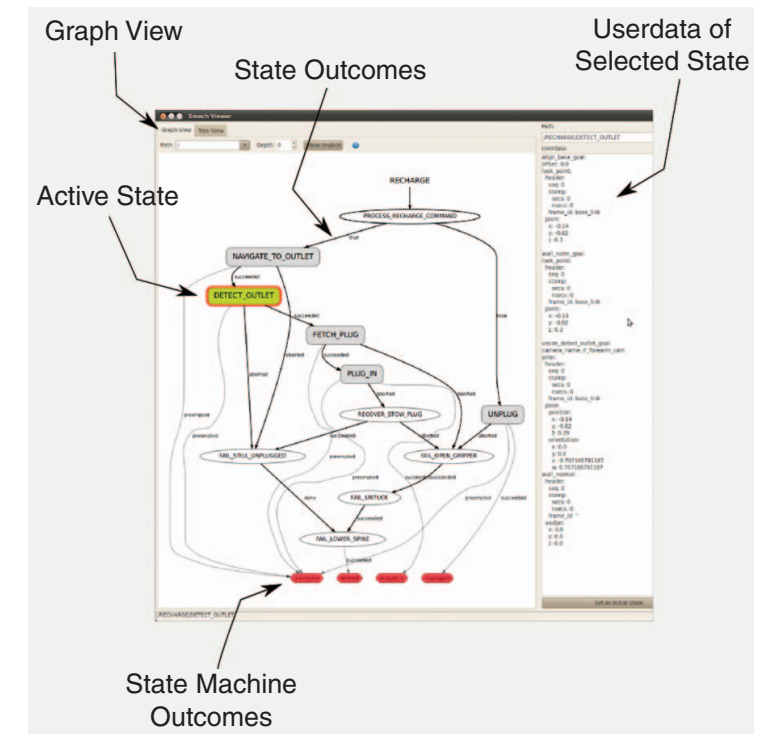
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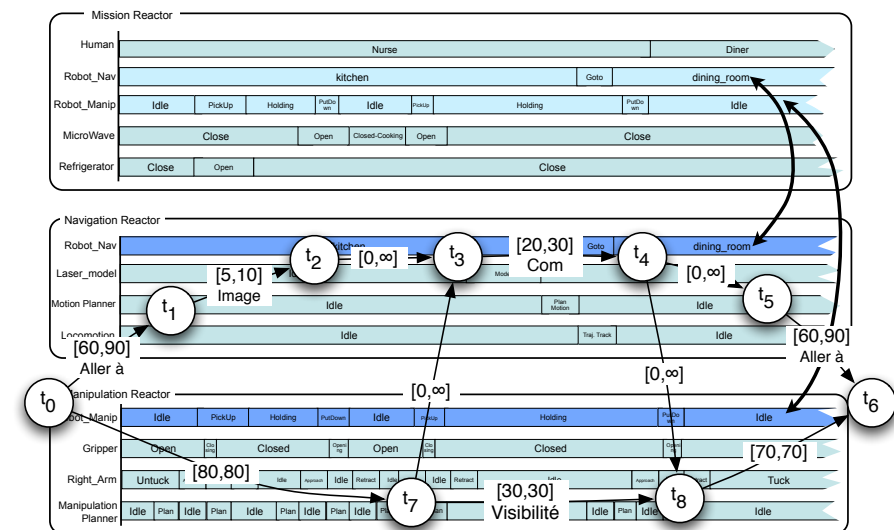
[Smach (ROS)]

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[Smach (ROS)]



[T-Rex / IDEA]

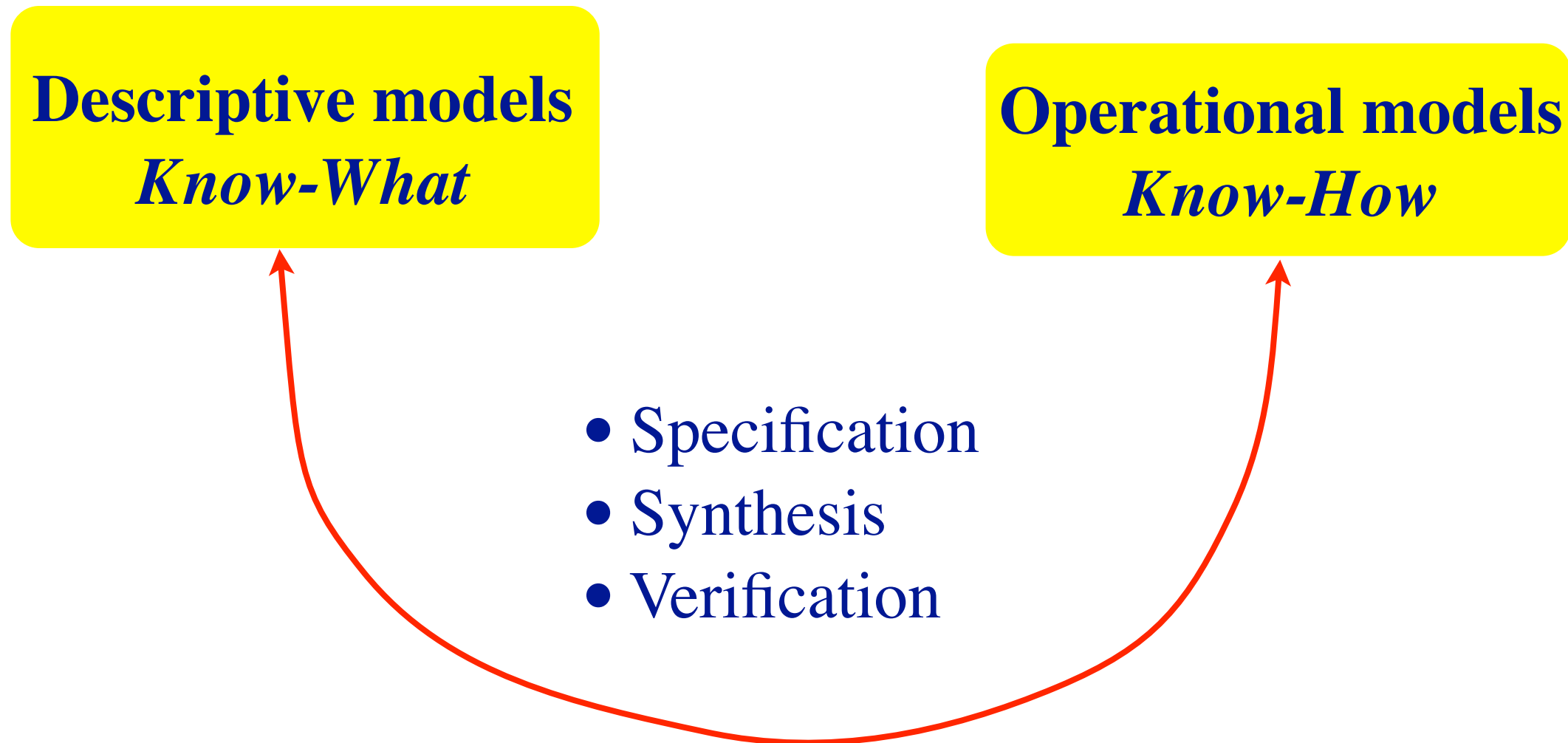
Approaches

Approaches	Systems
Direct	STRIPS/Planex
Procedure	RAP
	PRS
	Cypress/CPEF
	TCA/TDL
	XFRM/RPL/SRP
Petri-Net	IMRS
	Procosa
	Petri-Net Plans
Automata Graph	FSA
	PLEXIL
	RMPL
	SMACH
Logic	Golex
	ReadyLog
CSP	IxTeT
	RMPL
	IDEA/T-ReX
	Casper
MDP	PEARL
	K9 MDP
	Robel
	Ressac

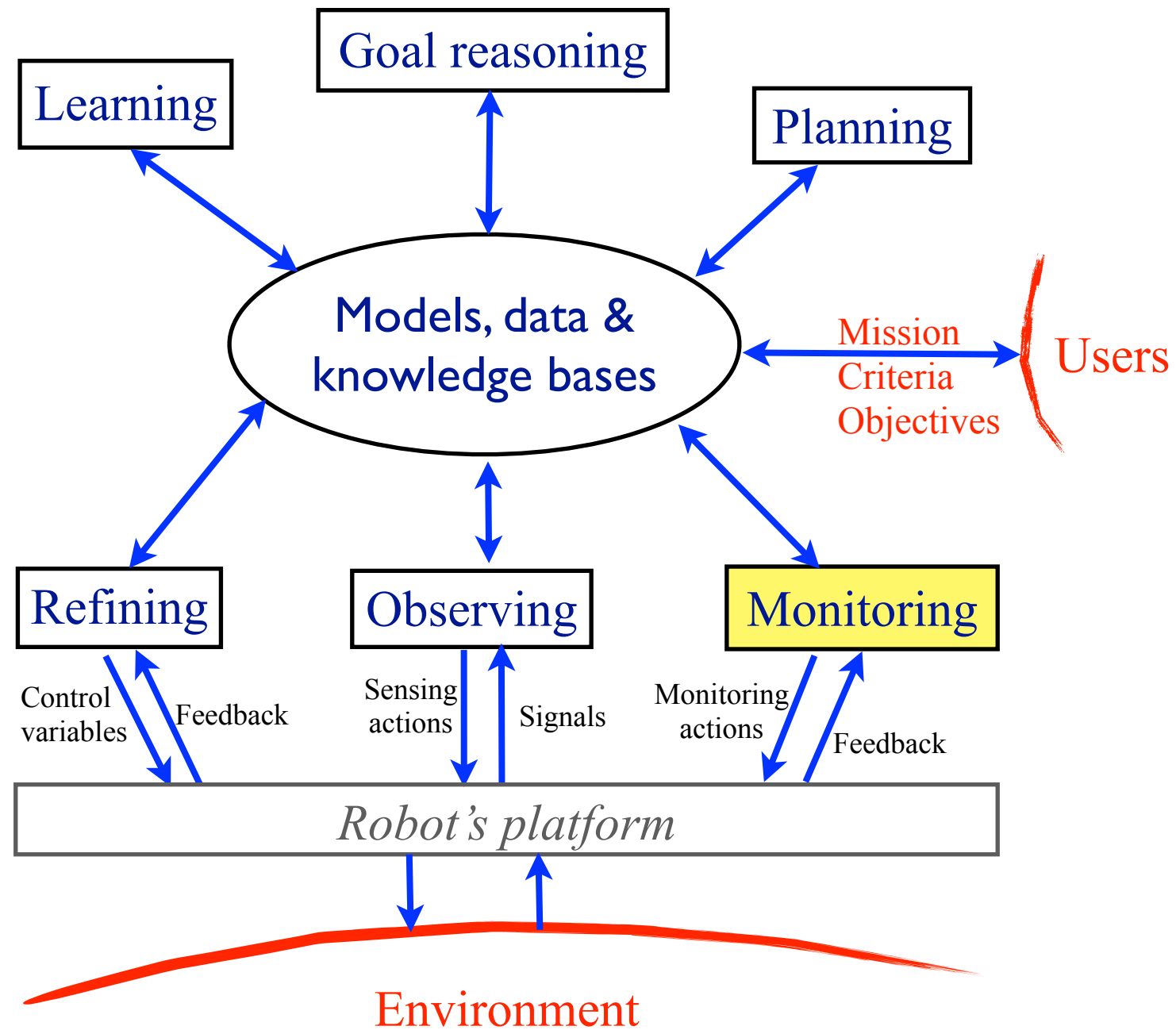
Approaches

Approaches	Systems	Functions					Knowledge Representation			
		<i>Refinement</i>	<i>Instantiation</i>	<i>Time handling</i>	<i>Nondeterminism</i>	<i>Repair</i>	<i>Hand-Written</i>	<i>Model Based</i>	<i>V&V</i>	<i>Similar to Planning KR</i>
Direct	STRIPS/Planex					X	X			
Procedure	RAP	X	X				X			
	PRS	X	X				X			
	Cypress/CPEF	X	X				X			ACT
	TCA/TDL	X	X	X			X			
	XFRM/RPL/SRP	X	X			X	X			X
Petri-Net	IMRS	X		coordination			X	X	X	
	Procosa	X		coordination			X	X	X	
	Petri-Net Plans	X		coordination			X	X	X	
Automata Graph	FSA	X	X	X			X	X		Link
	PLEXIL	X	X	X			X	X		
	RMPL	X	X	X			X	X		
	SMACH	X	X	X			X	X		
Logic	Golex	X	X				X		X	Same lang. not same model
	ReadyLog	X	X				X		X	
CSP	IxTeT	X	X	X		X		X		X
	RMPL	X	X	X	X	X		X		X
	IDEA/T-ReX	X	X	X		X		X		X
	Casper	X	X	X		X		X		X
MDP	PEARL				X					
	K9 MDP				X					
	Robel				X					
	Ressac				X					

Refining abstract steps into skills



Actor's deliberation functions



► Functions

- Survey actor's predictions
in plans, skills and environment models
- Detect discrepancies = predictions - observations
- Explain and diagnose discrepancies
- Recover: trigger first reactions and repair actions

Approaches

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- ▶ Hand-written procedures to monitor applicability/maintenance conditions

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- ▶ Synthesize additional invariants from extended planning problems which guarantees the plan execution, e.g., TALPlan

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- ▶ Model checking execution traces with LTL

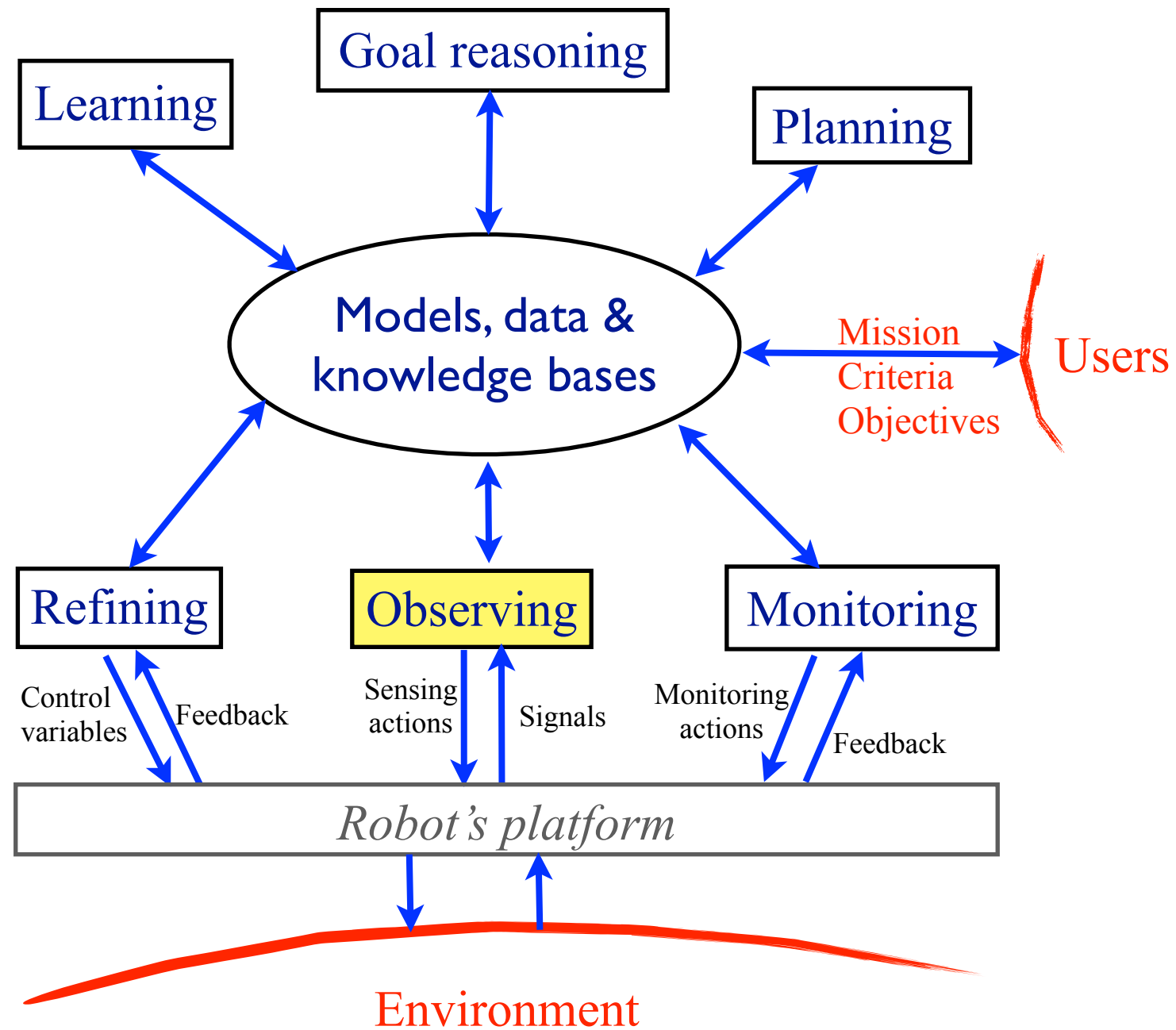
Approaches

- ▶ Hand-written procedures to monitor applicability/maintenance conditions
- ▶ Model-based supervision and diagnosis, e.g., DS1 more proprioceptive than exteroceptive
- ▶ Monitor plan invariants
- ▶ Synthesize additional invariants from extended planning problems which guarantees the plan execution, e.g., TALPlan
- ▶ Model checking execution traces with LTL
- ▶ Constraint Based Automata + Control programs, e.g., RMPL

Approaches

	<i>Knowledge Representation</i>	<i>Search & Algorithm</i>	<i>Link with other Deliberative Functions</i>
Planex	Triangular table		with planning and exec
Livingstone	Qualitative model Transition automata with probabilities and costs	ITMS with conflict directed best-first search	
RMPL	Hierarchical constraint-based automata	Reactive programming	with planning
[Fraser et al., 2005]	Logical invariant	Logical satisfiability	with planning
[Fichtner et al., 2003]	Fluent Calculus	Prioritized non monotonic default logic	
[Lamine and Kabanza, 2002]	Linear Temporal Logic	Delayed formula progression	same technique used for planning
[Pettersson et al., 2003]	Neural net		
SKEMon	Description Logic & Bayesian Belief		
TALplanner	Temporal Action Logics	Formula progression	with planning and observing

Actor's deliberation functions



► Role

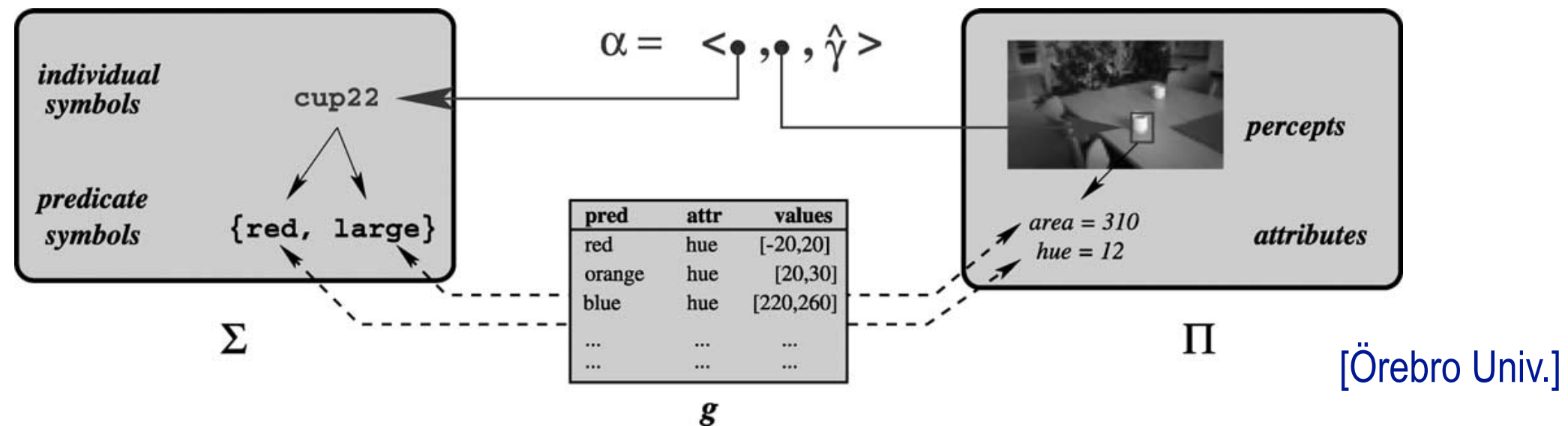
- Process signals needed in closed loop servoing
- Detect and structure environment features
Recognize, categorize,
Link signals to symbols: anchoring
- Recognize situations and plans in observed sequences of events

► Bottom-up to from signal to symbols

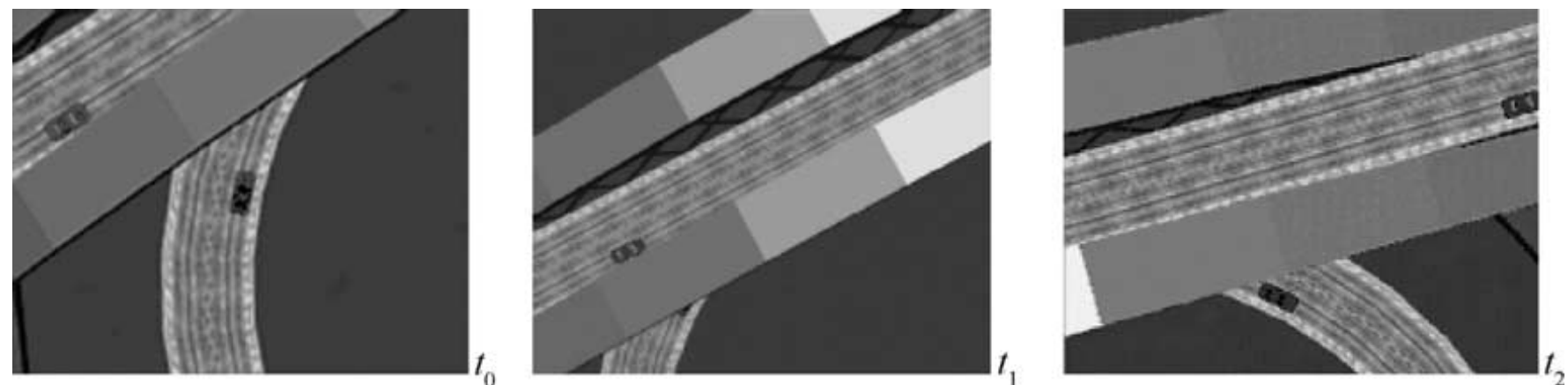
► Top-down to focus attention and trigger observation actions

► Anchoring problem

- Relate perceptual data and symbolic attribute corresponding to the same physical object

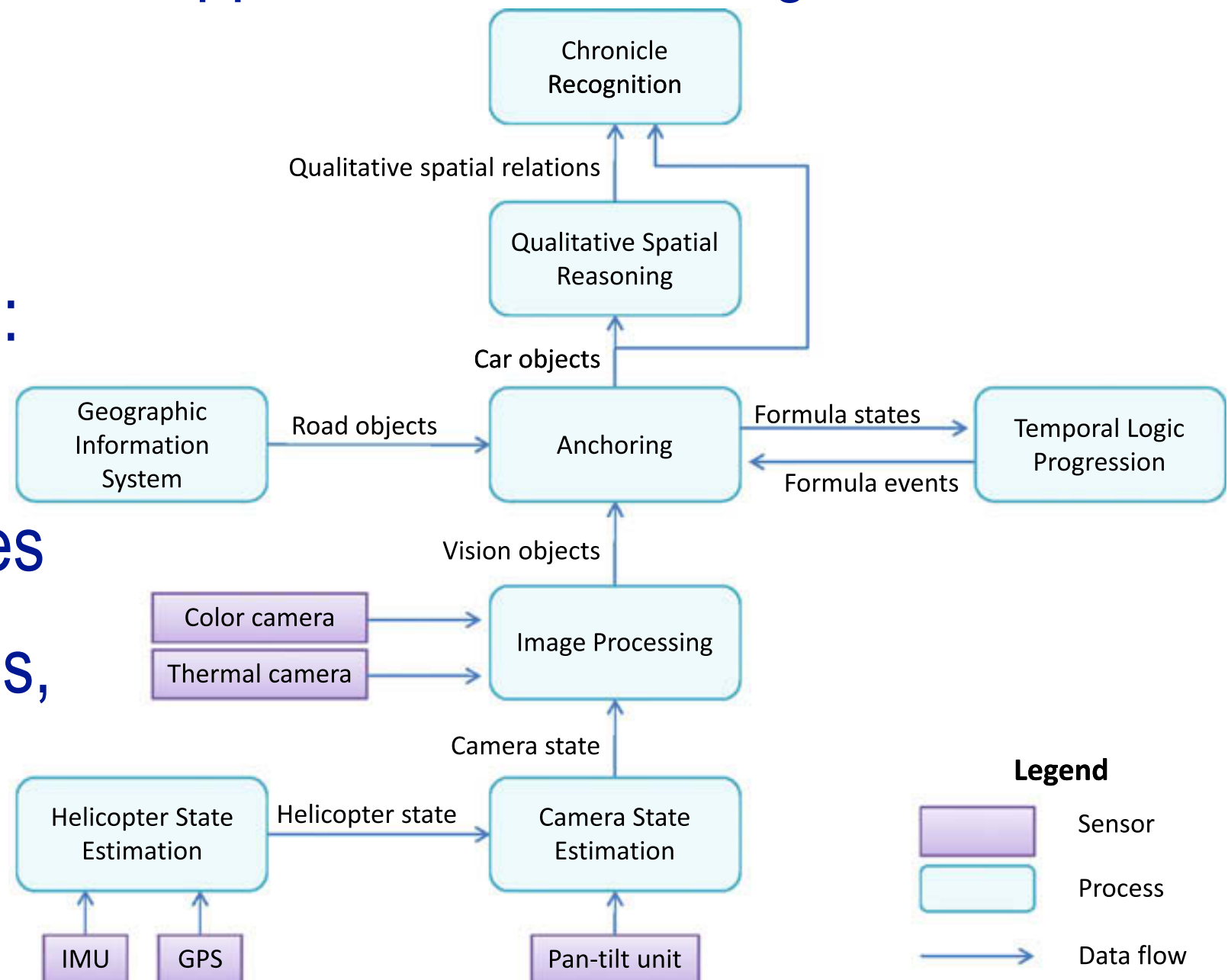


- Track anchors overtime and refine/revise hypothesis



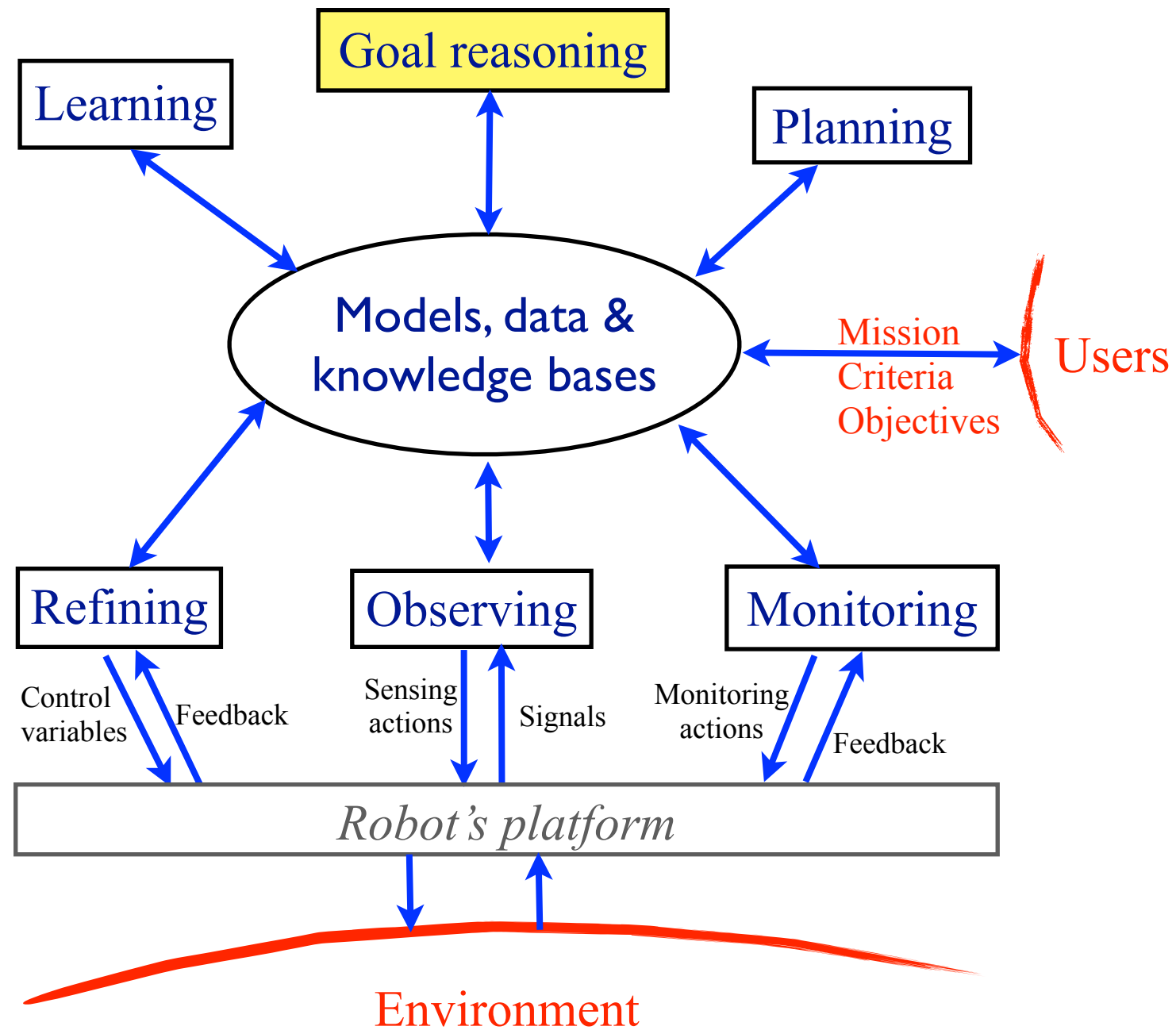
Perception Engine

- ▶ Comprehensive and coherent approach for observing
- ▶ Data flow architecture
- ▶ Stream-based formalism on perception processes:
 - primitive, refinement, configuration processes
 - policies over processes, temporal constraints
- ▶ Integration to planning and monitoring,
- ▶ Opens V&V perspectives



[Dyknow, Linköping U.]

Actor's deliberation functions



Goal Reasoning

- ▶ Higher level monitoring *wrt* objectives, criteria and constraints

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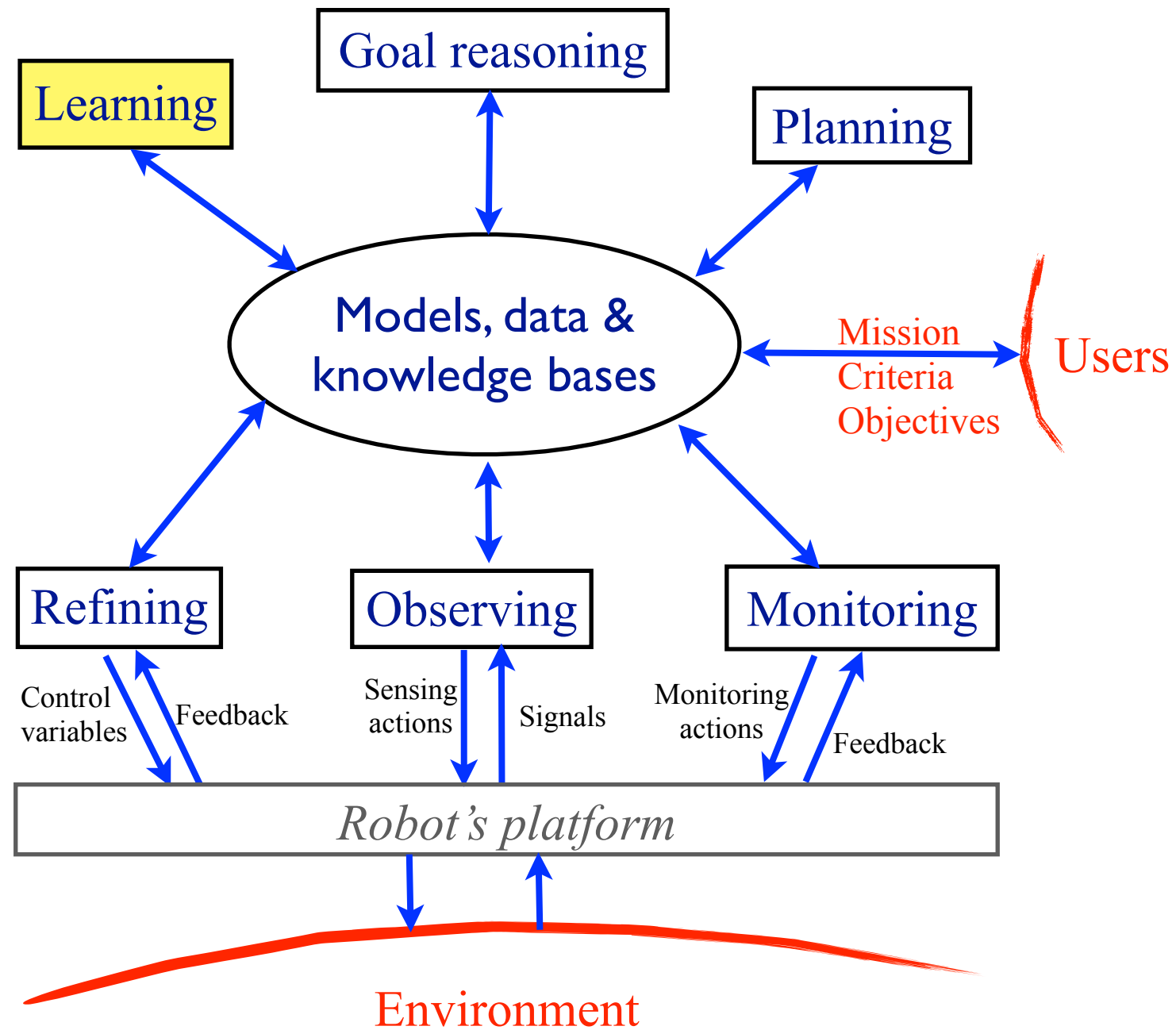
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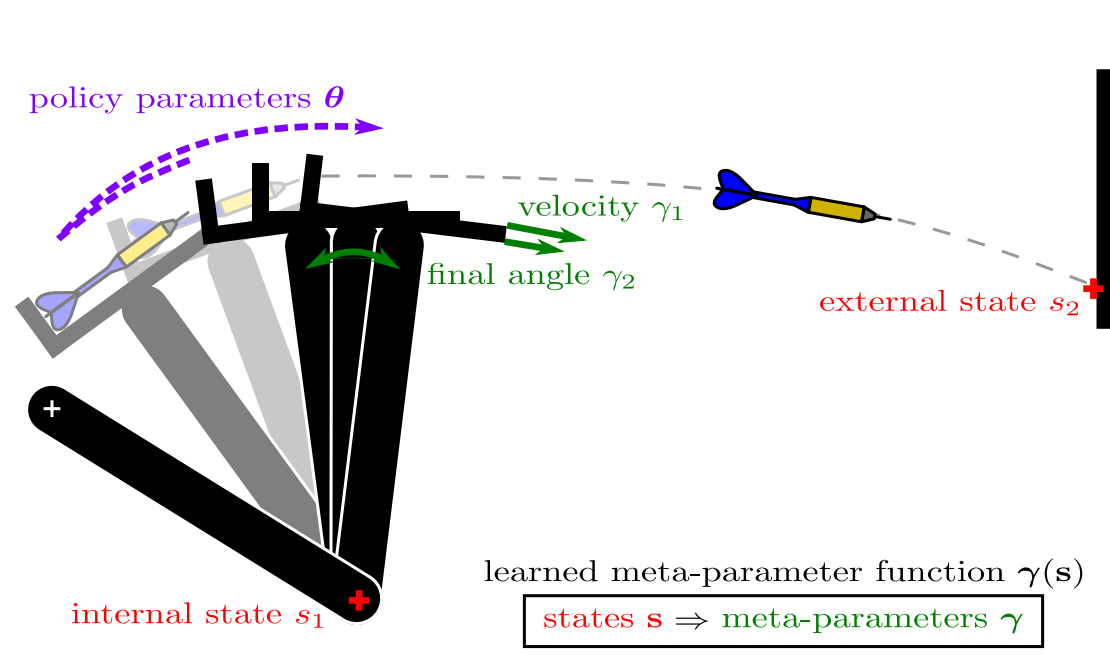
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 - Decision theory: tradeoff between conflicting goals
 - Explicit choice
 - ▶ Found in large systems, e.g., CPEF, DS1
- Function often embedded in acting/monitoring/planning

Actor's deliberation functions

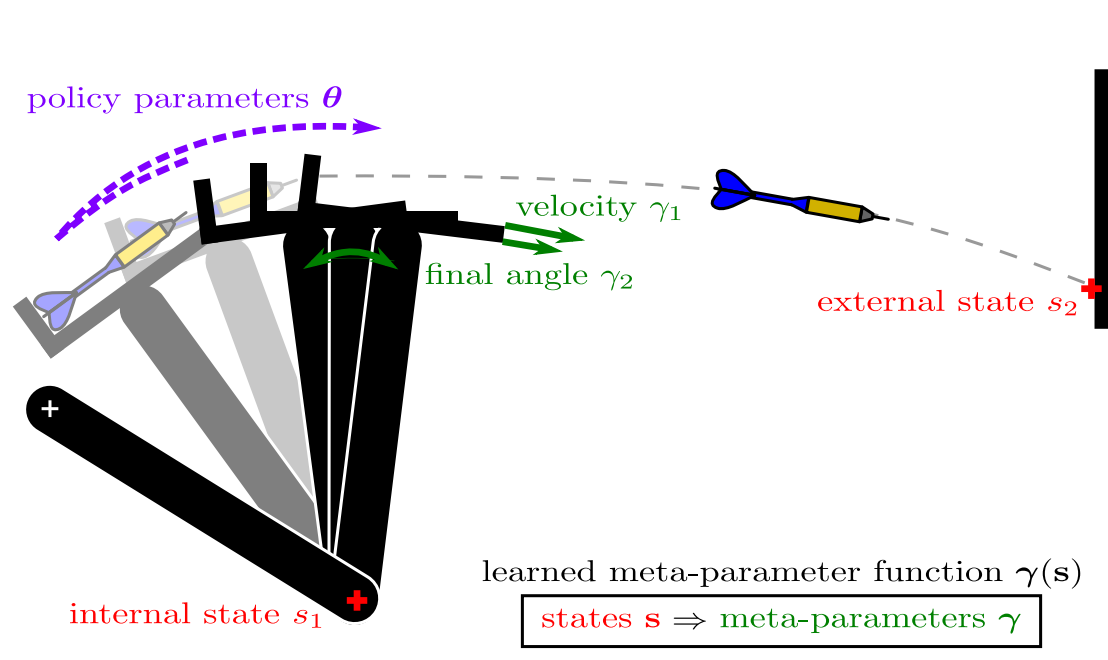


Reinforcement Learning of sensory motor functions

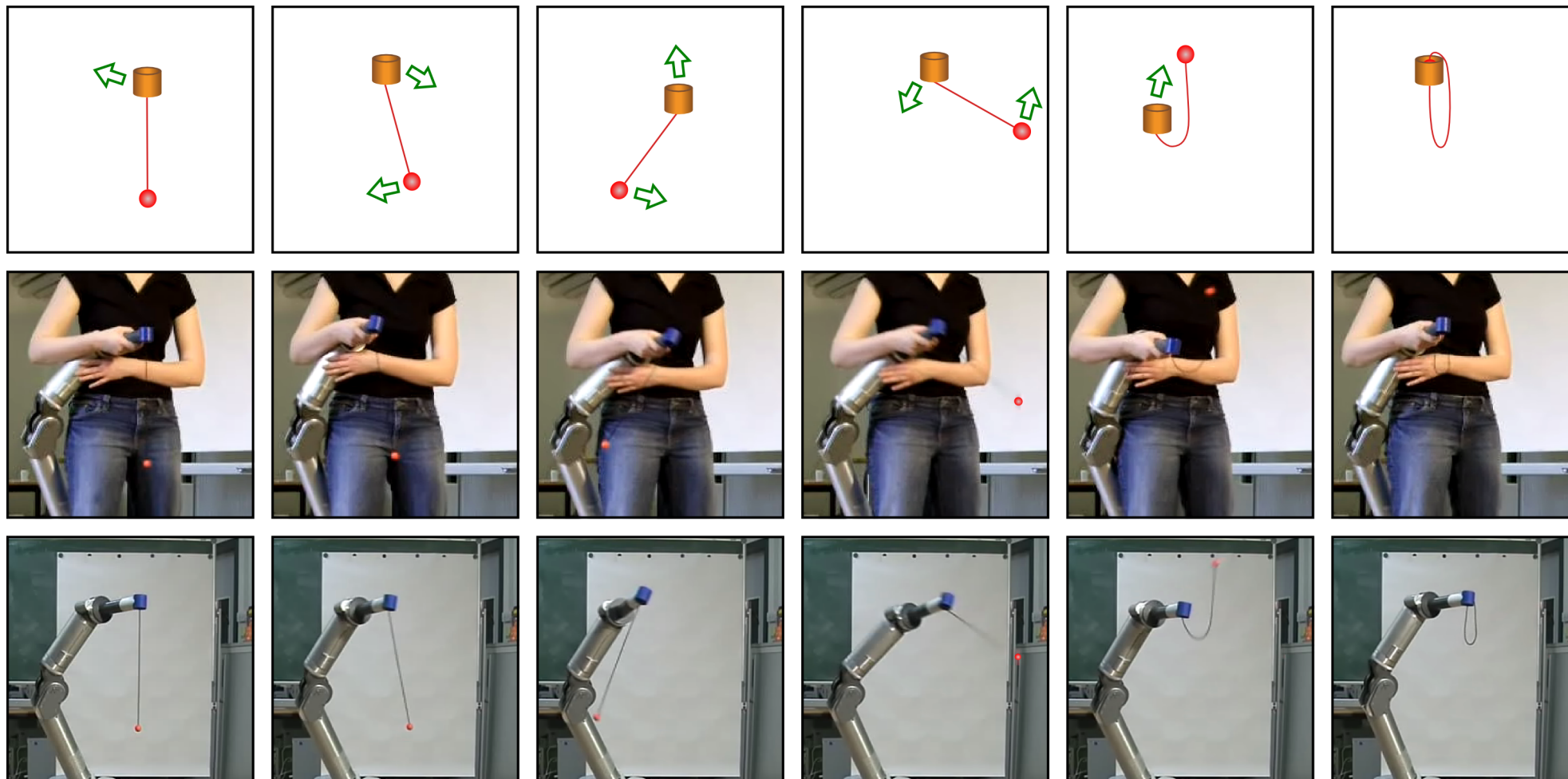


[MPI Bio-Cybernetics, Tübingen]

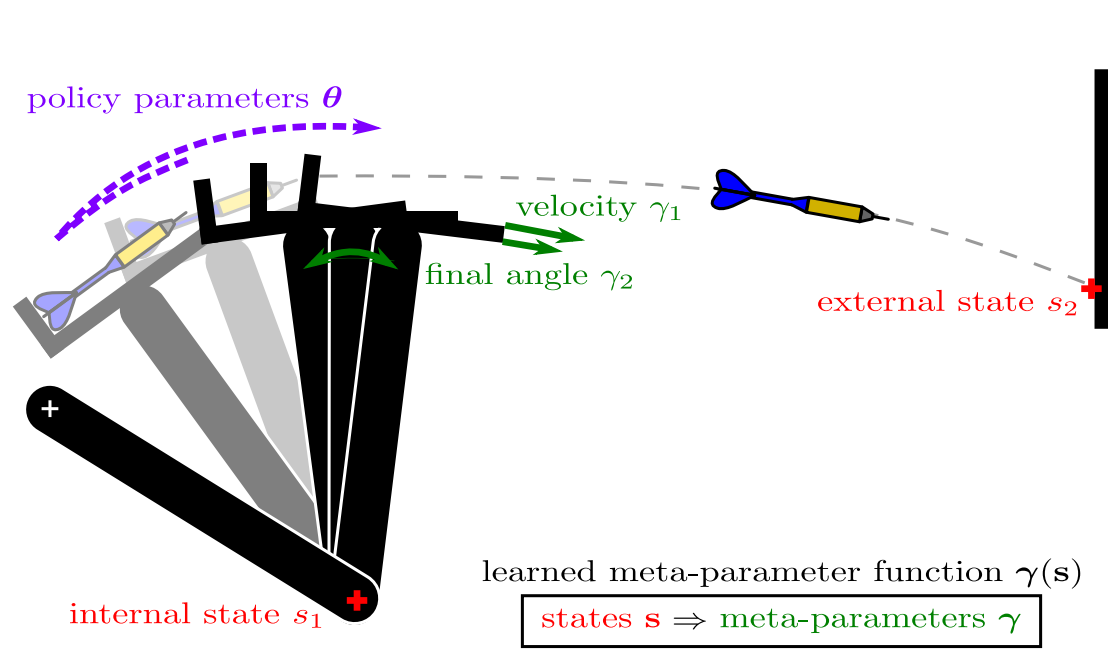
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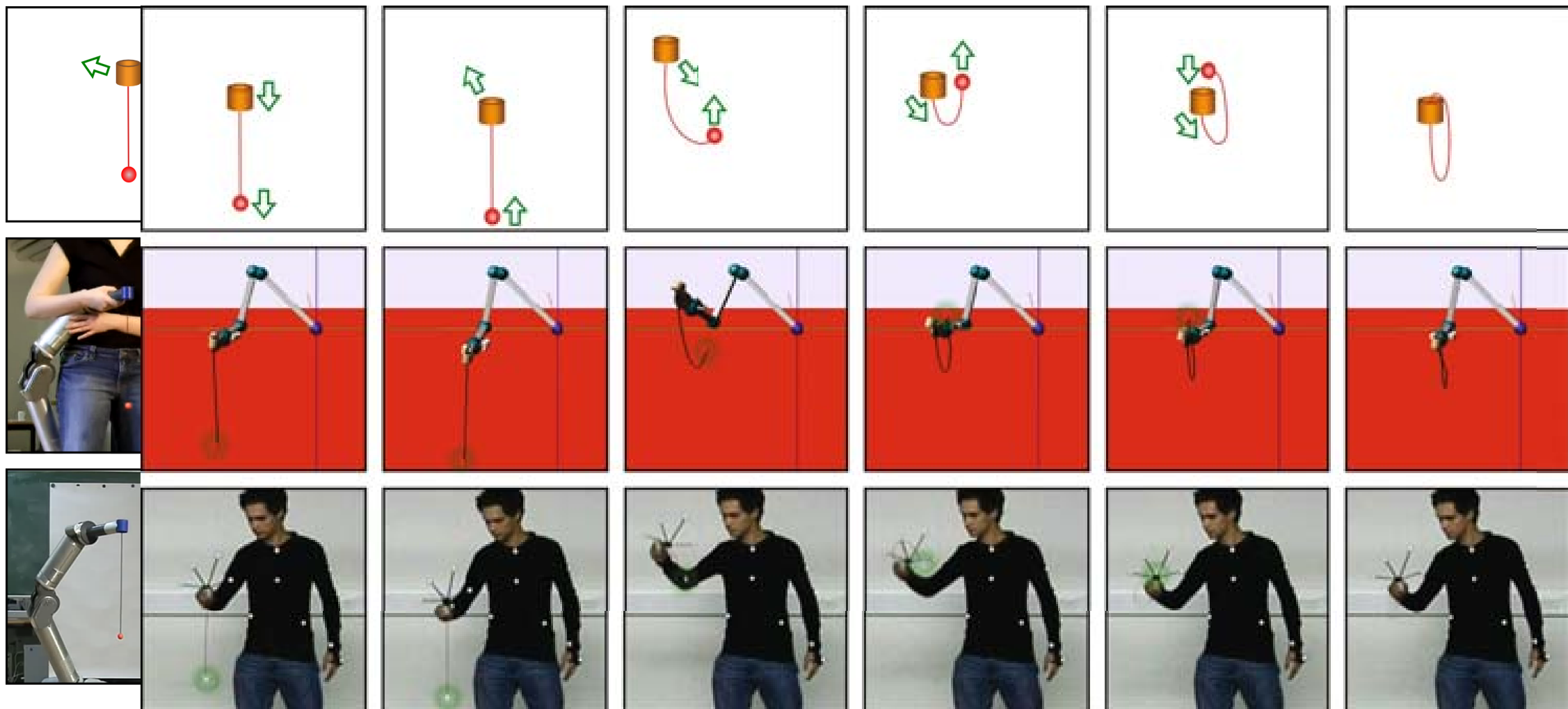
[MPI Bio-Cybernetics, Tübingen]



Reinforcement Learning of sensory motor functions



[MPI Bio-Cybernetics, Tübingen]



Helicopter Aerobatics Apprenticeship Learning

[U. Stanford]

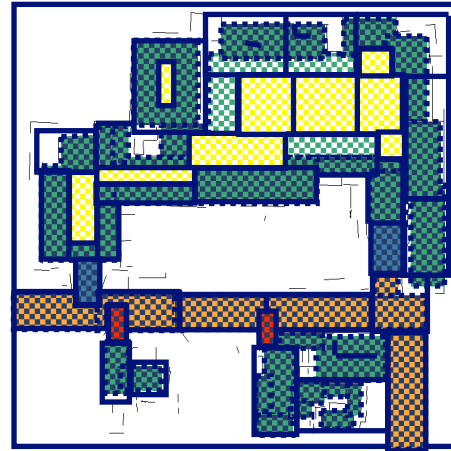
Aerobatics Apprenticeship Learning

Simple linear rigid dynamic models of helicopter

- ▶ Learn dynamic models, *one for each type of maneuver*
 - Regression from teacher's demonstrations
 - Improvement by reinforcement learning in autonomous flight
- ▶ Learn reference trajectories, one for each aerobatic figure
 - *Expectation-Maximization* on teacher's demonstrations
 - Temporal alignment and optimization
- ▶ Learn controllers, one for each aerobatic figure
 - *Differential dynamic programming* continuous MDPs solved by iterative approximation of receding horizon LQR problems

Skill selection learning

Navigation goal



Markov Decision Process

- Optimal plan
- Receding Horizon Control

S2

Start

Stop

Query resource

Resource allocation

Resource Manager

Suspend
Resume

Extensible set of skills

Skill 1

Skill 2

Skill 3

Skill 4

...

action

action

action

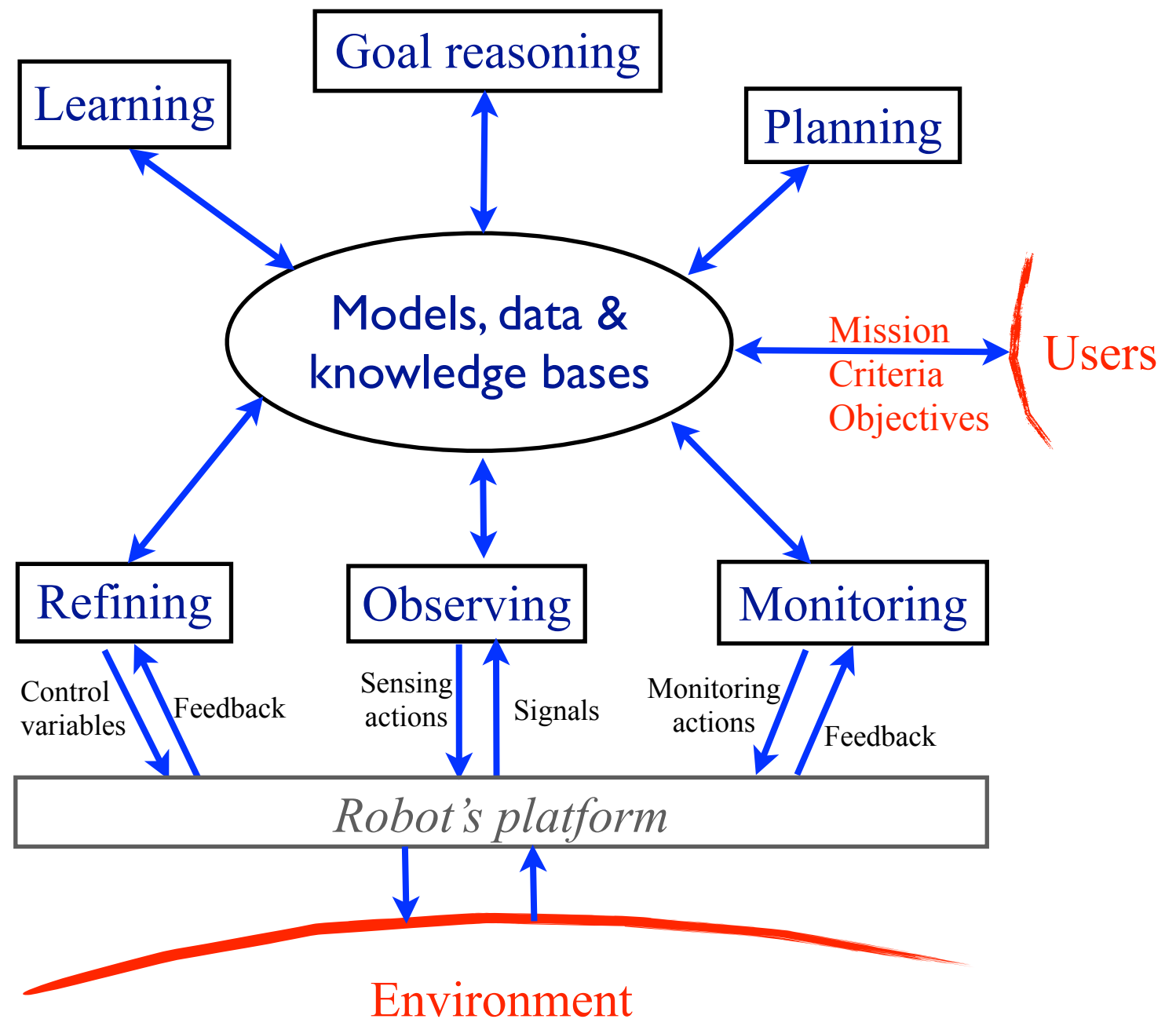
action

Reactive Control System

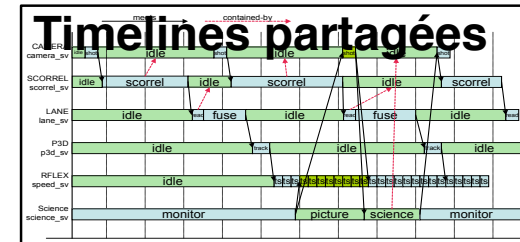
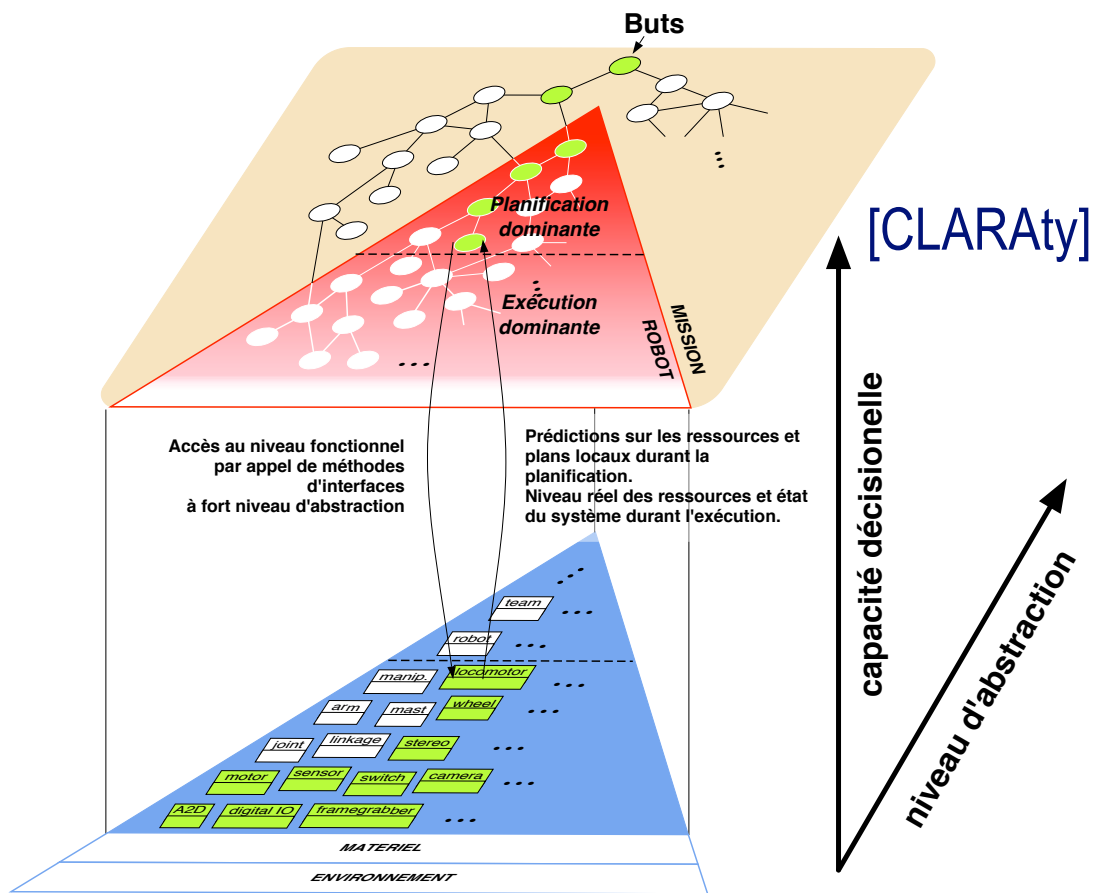
Reports

Sensory-motor commands

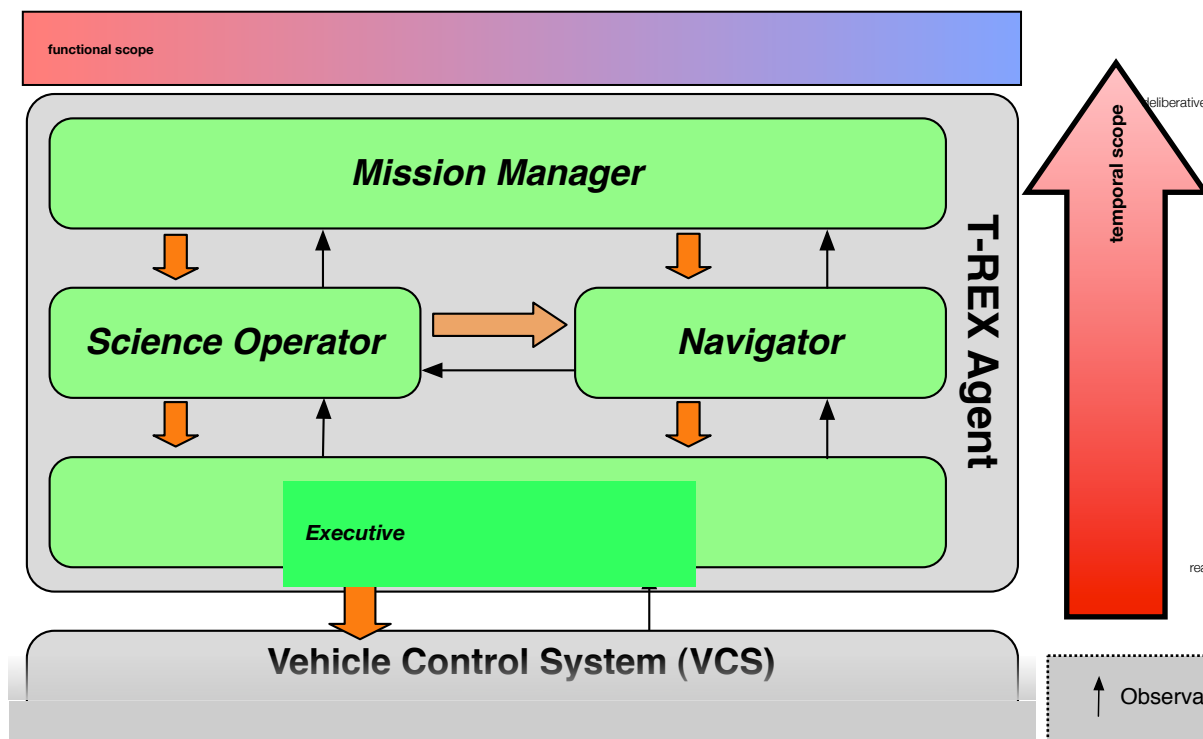
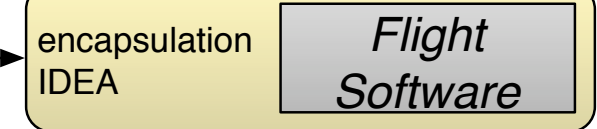
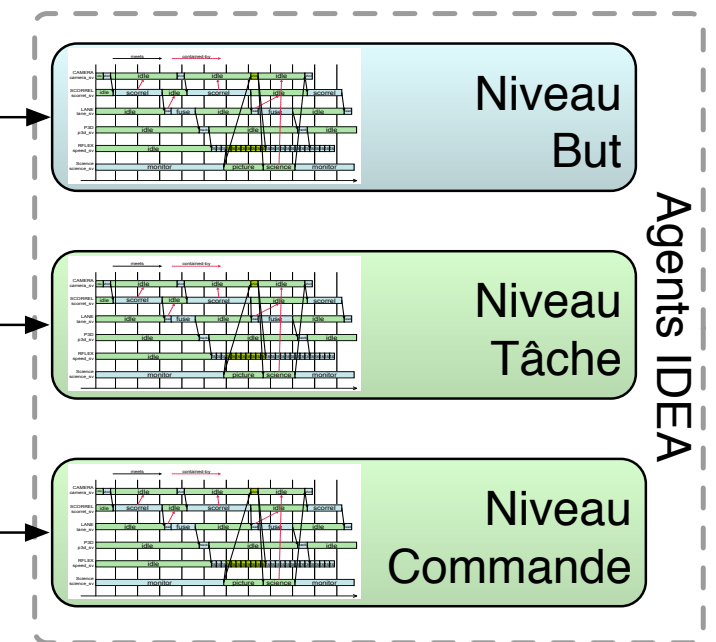
Actor's deliberation functions



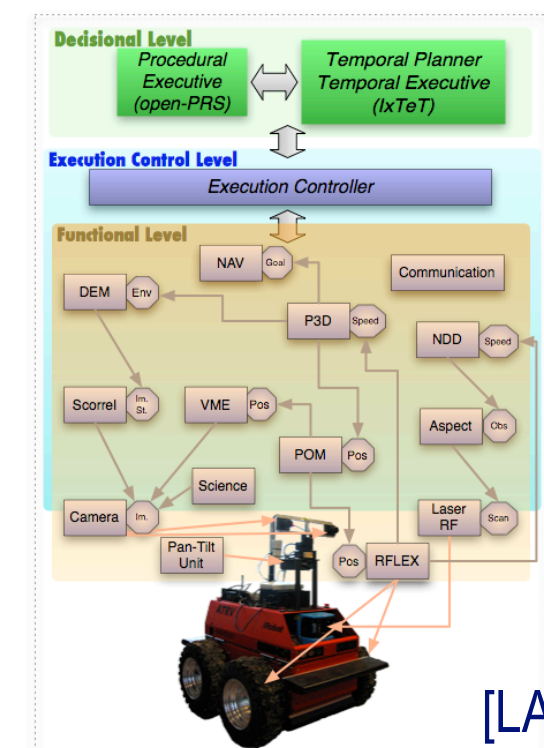
Integration



[IDEA]

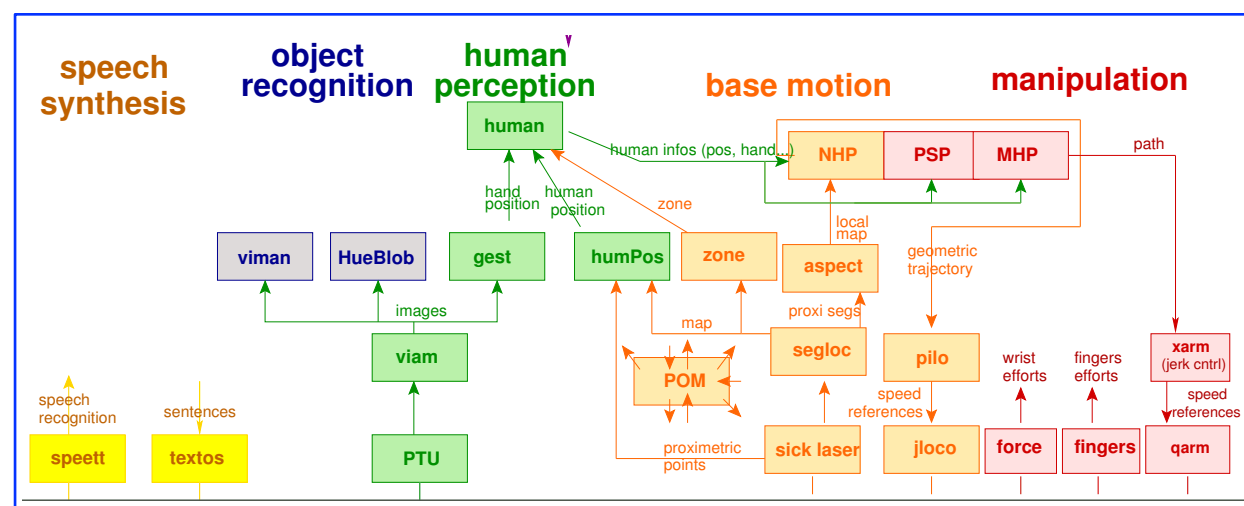
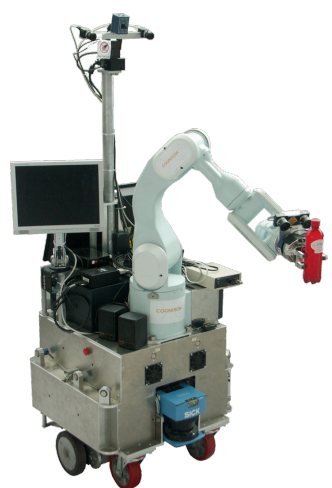
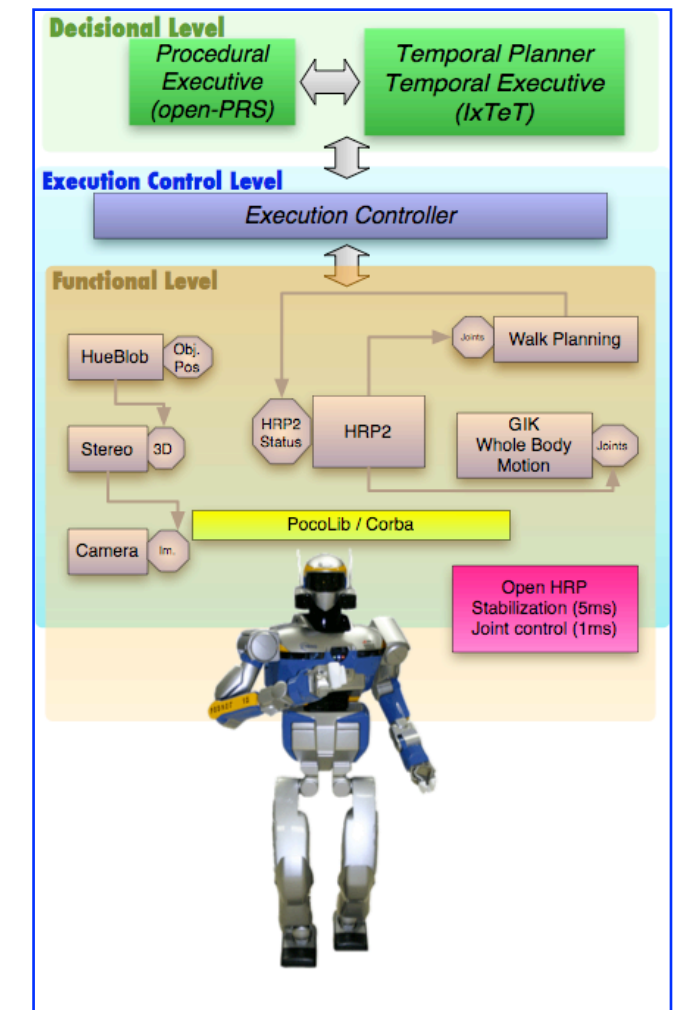
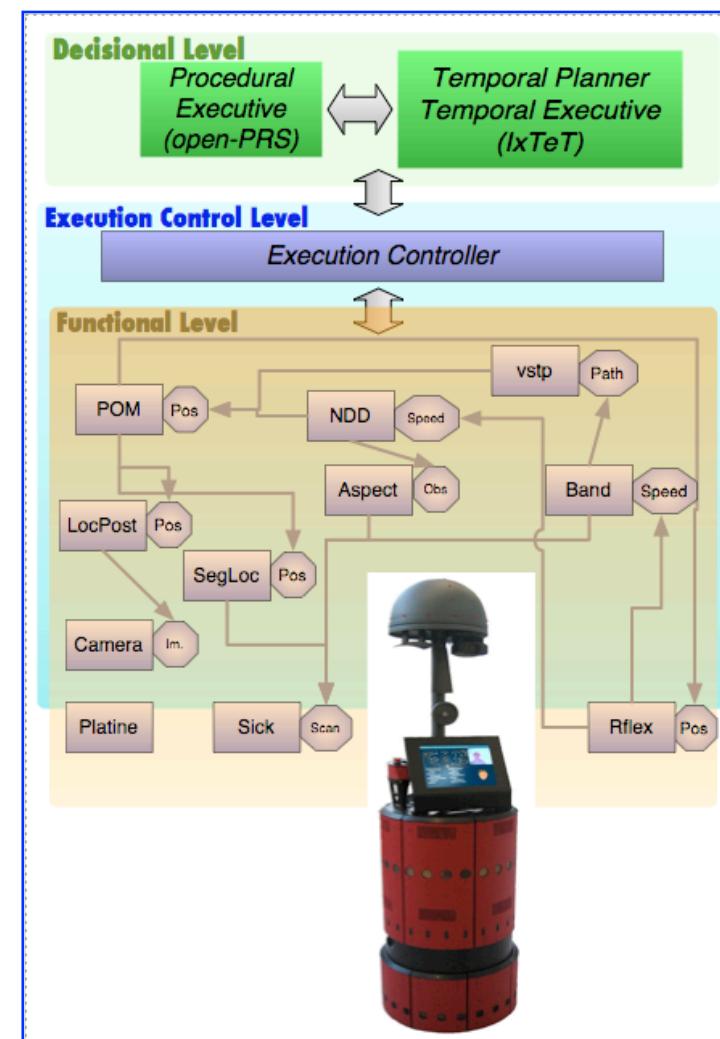
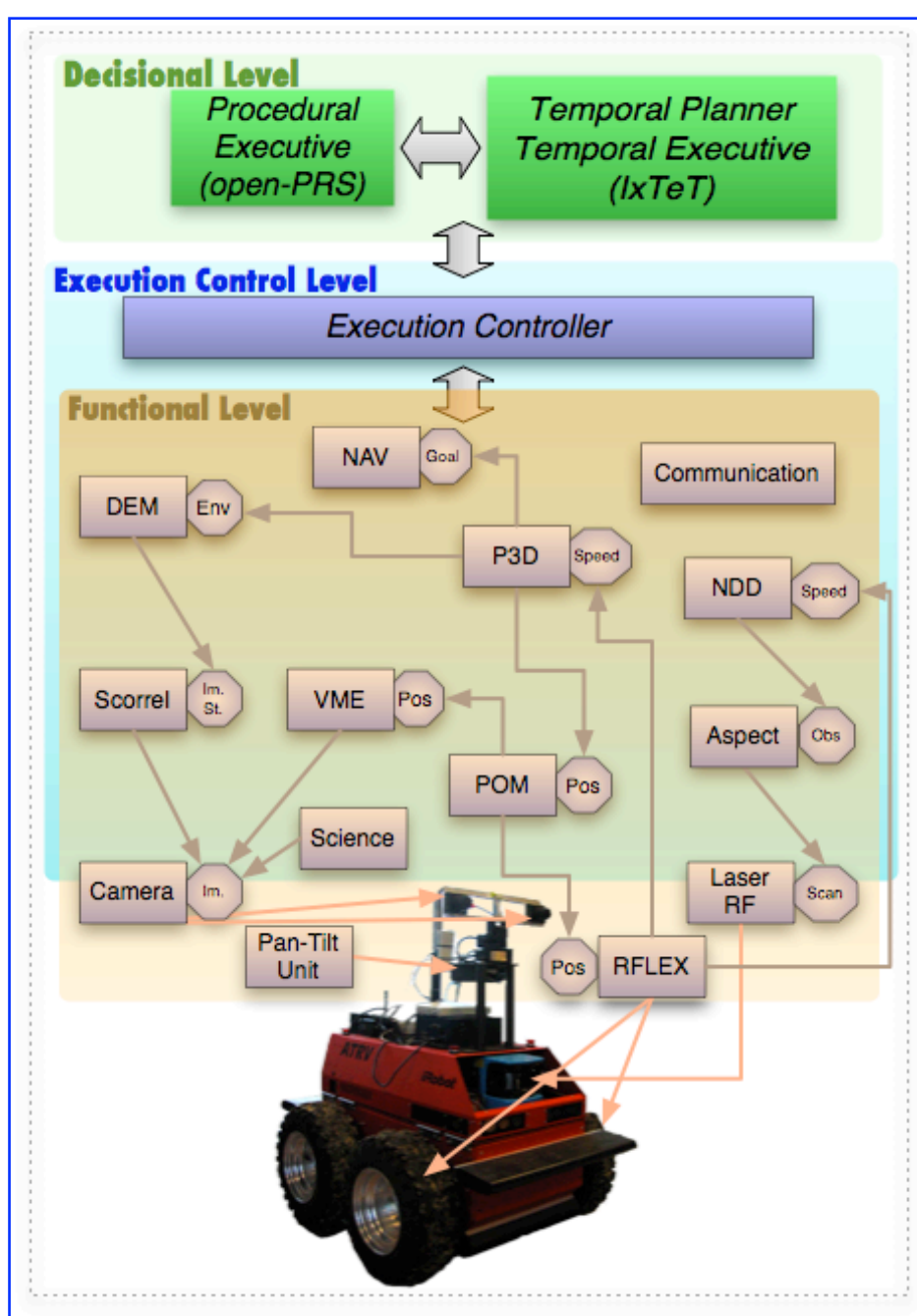


[T-ReX]

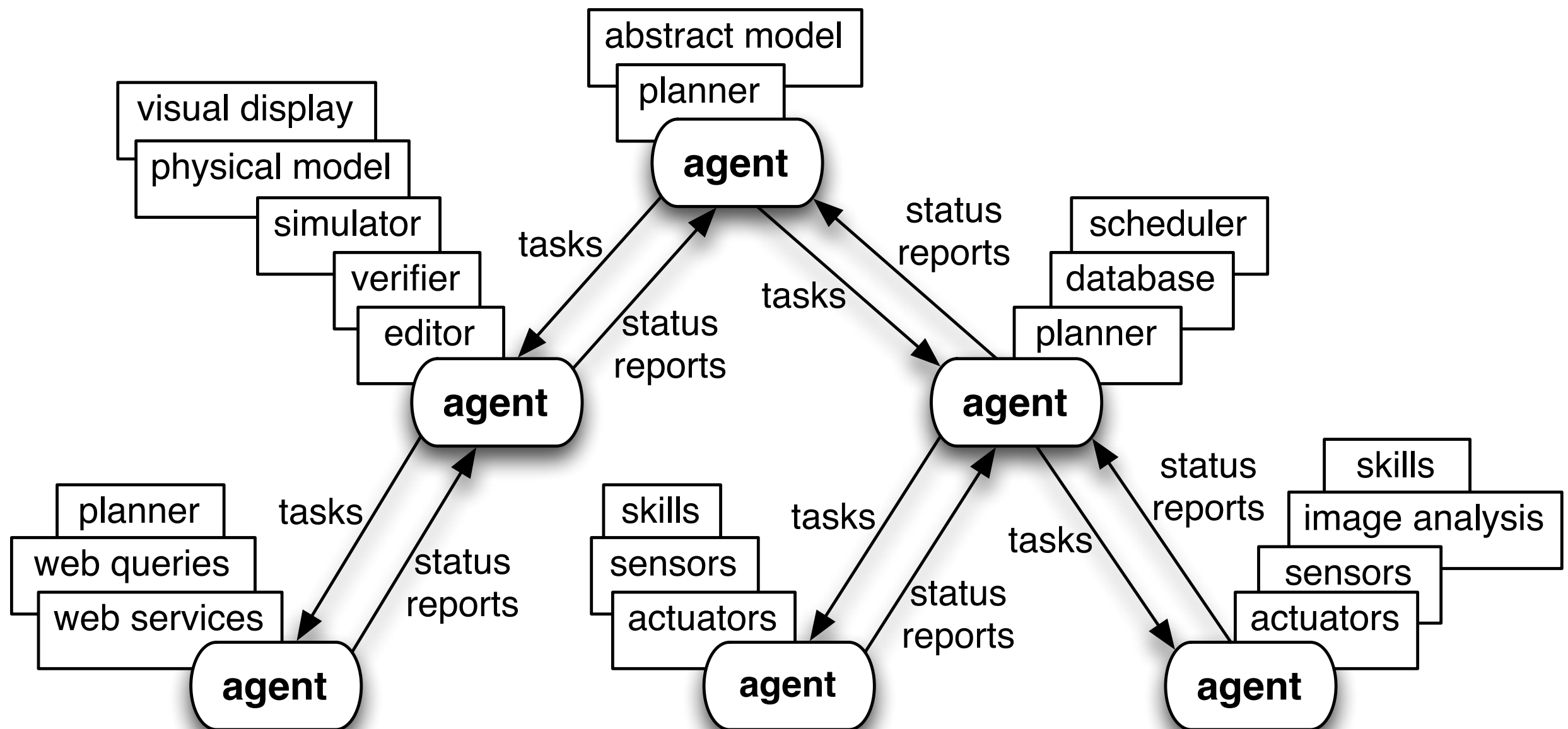


[LAAS] 44

Integration



- Organizational principle: *actor as a hierarchy of agents*



✓ Motivations

✓ Deliberation functions

- Planning
- Refining
- Monitoring
- Perceiving
- Goal reasoning
- Learning
- Integration

► Research Challenges

- Representation
- Model acquisition & Verification
- Synthesis & Refinement
- Monitoring and Goal reasoning
- Integration

Representation Challenges

► ***Descriptive models***

- Map known representations: PDDLx, ANML, RDDDL, \mathcal{K}
 - Where in the actors hierarchy
 - Link to monitoring, sensing, control
 - Suitability for online plan repair
 - Criticality issues
- Link to open domain representations: ontologies/DL, e.g., RoboEarth, OMRKF, ORO, RACE

- ***Operational models***: procedural, automata and graphical
- Relationships between descriptive and operational models
- Simulation and sampling techniques

Model Acquisition, Learning and Verification

- ▶ Tools for the specification of descriptive and operational models
- ▶ Learning to acquire or improve these models
 - Reinforcement Learning: hierarchical, relation RL, factored MDP
 - Learning from demonstration: teleoperation, external observation
- ▶ Verification
 - Hierarchy of actors: of the consistency of their models
 - Heterogeneity of representations
 - Program verification techniques
 - Model checking

Synthesis and refinement

- ▶ Online plan synthesis, extension and repair, while acting
- ▶ Online skill selection and adaptation
- ▶ Integrate temporal dimensions:
 - Time in reasoning about a peculiar resource
 - Time as a computational resource for reasoning
 - Real-time constraints on acting and deliberation
- ▶ Planning with sensing and information gathering actions
- ▶ Integrate risk and criticality considerations to plan horizon and optimization issues

Monitoring and goal reasoning

- ▶ Derive monitoring conditions from descriptive and operational models
- ▶ Focus of attention mechanism and link to perceiving for acquiring information needed for monitoring
- ▶ Model-based diagnosis for the robot-environment interactions
- ▶ Recovery actions and link to criticality analysis issues
- ▶ Qualify current goals with respect of longer term objectives and motivations, express reservations and conditions to be monitored
- ▶ How to synthesize new goals for current mission
- ▶ Map monitoring functions to the actor's hierarchy

- ▶ How to organize actors hierarchy
- ▶ Static, i.e., mapped to the robot architecture, or dynamic
- ▶ Actor's enablers, including executors
- ▶ Concurrency of actors
- ▶ Temporal constraints

Other Deliberation challenges in robotics

- ▶ Observing the environment semantics
- ▶ Interacting
- ▶ Learning
 - Models of the robot and the environment
 - Categories
 - Functions, skills and behaviors
- ▶ Architecture
 - Specification
 - Robust adaptation

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Actor's view of deliberation: numerous challenges
Planning is just the tip of the iceberg