



16.410 and 16.412: Principles of Autonomy and Decision Making

Prof Brian Williams, Prof Emilio Frazzoli and Sertac Karaman
September, 8th, 2010



Assignments

- Homework:
 - Class signup, return at end of class;
 - Problem Set #1: Java warm up
Out Today,
Due next Wednesday, September 15th
- Reading:
 - Today: *[AIMA] Ch. 2, [JINS] Ch. 1-3,5*
 - Monday: search *[AIMA] Ch. 3.1-4*



Outline

1. Trends in Computing
2. Examples of autonomous systems - Williams
3. Autonomous systems architectures
4. Principles = modeling + inference + search
5. More examples – Frazzoli
6. Course logistics and schedule
7. Projects and programming

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Human Brain

- 100 Billion neurons
- On average, connected to 1 K others
- Neurons are slow. Firing rates < 100 Hz.
- Can be classified into
 - Sensory
 - Motor
 - Central
 - (reasoning, problem solving, language..)

Images of brain activity removed due to copyright restrictions.

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Trends in Biological and Machine Evolution - Moravec



- 1 neuron = 1000 instructions/sec
- Human brain then processes 10^{14} IPS
- 1 synapse = 1 byte of information
- Human brain then has 10^{14} bytes of storage
- In 2000, we had 10^9 IPS and 10^9 bytes on a desktop machine
- In 25 years, assuming Moore's law we obtain human level computing power



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Williams Research: Model-based Programming of Autonomous Systems



Robust, mission-directed agents:

1. Self-repairing agents
2. Agents that are agile
3. Science explorers

1. Self-Repairing Agents

- 7 year cruise
- ~ 150 - 300 ground operators
- ~ 1 billion \$
- 7 years to build

Affordable Missions

↓

- 150 million \$
- 2 year build
- 0 ground ops

Cassini Maps Titan

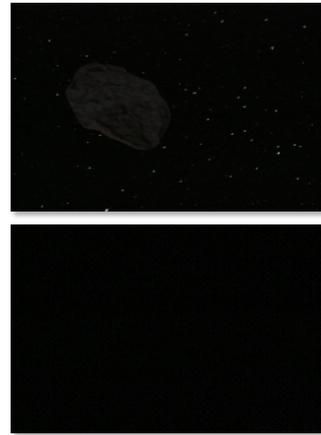
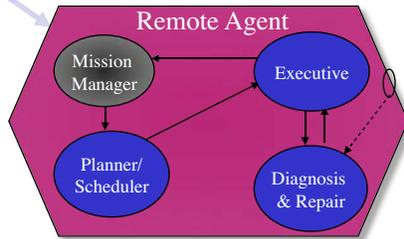
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Remote Agent on Deep Space One



Goals



1. Commanded by giving goals
2. Reasoned from commonsense models
3. Closed loop on goals

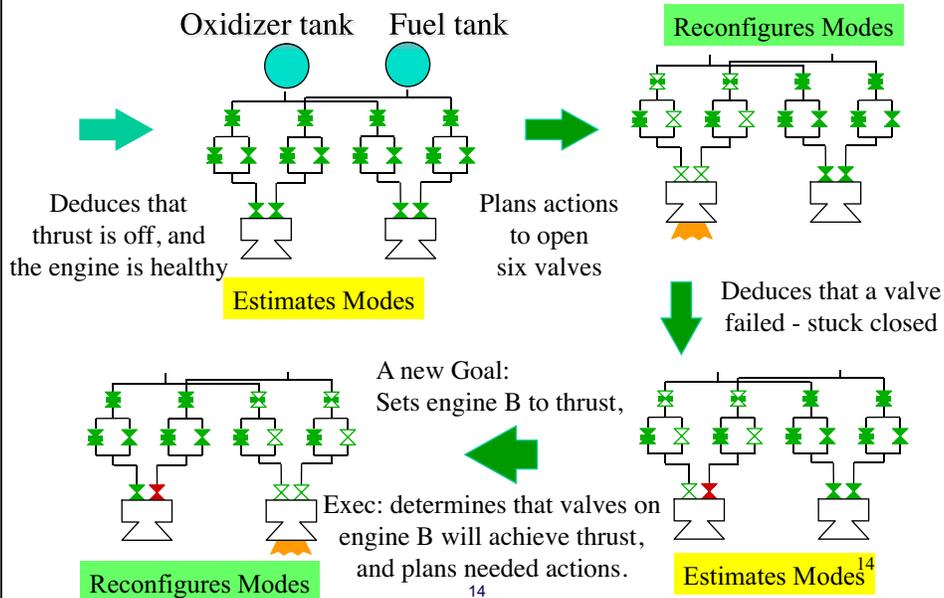
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[Williams & Nayak, AAI 95; Muscettola et al, AIJ 00]

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A Goal sets engine A state to thrusting, and the agent . . .



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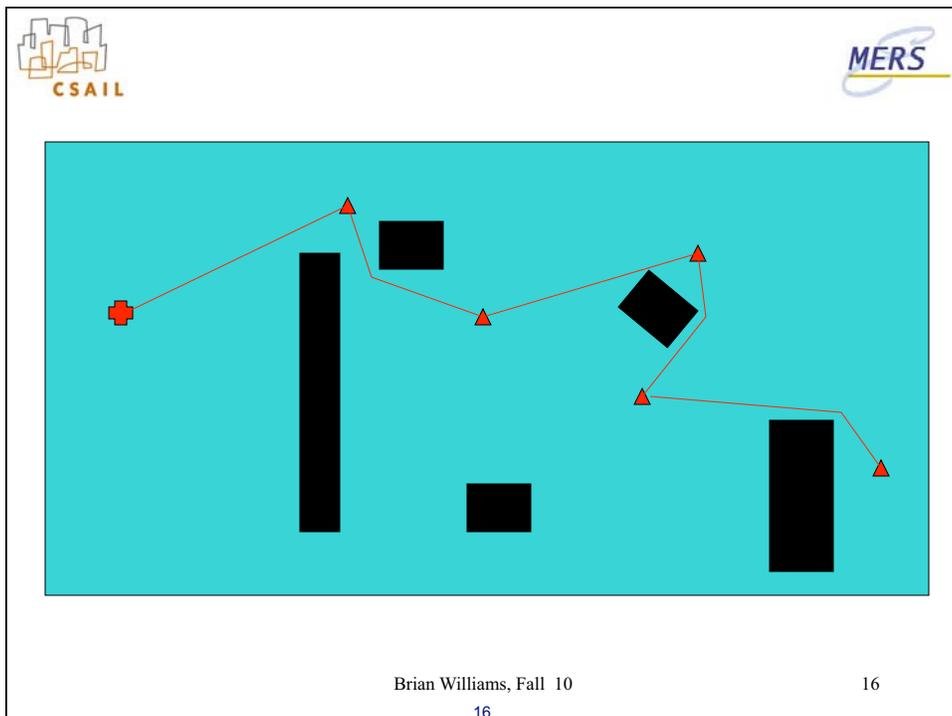
2. Agile Agents



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Image credit: NASA.



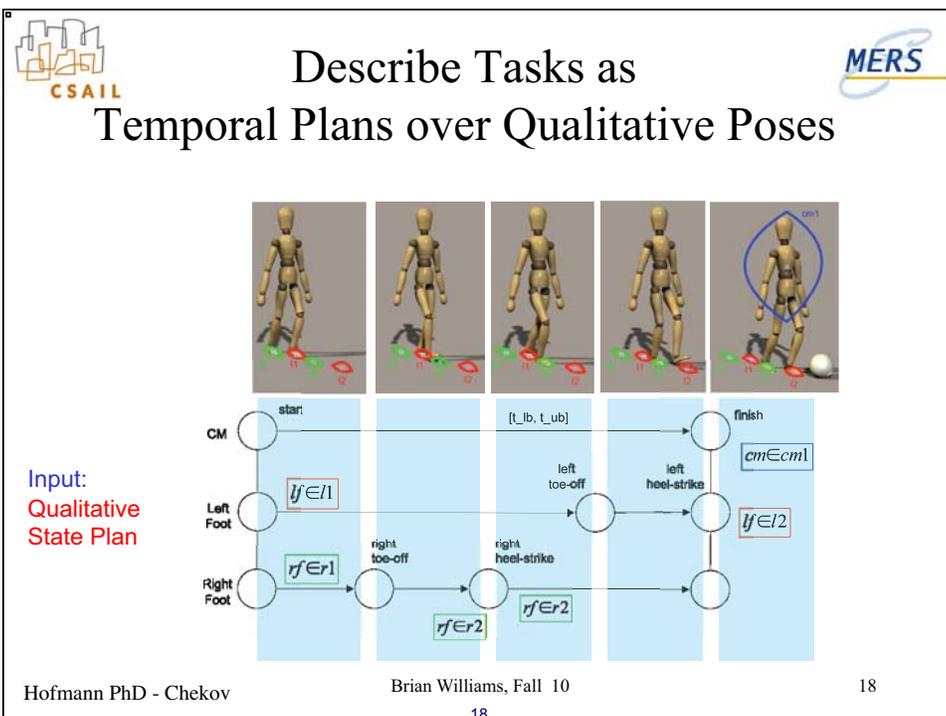
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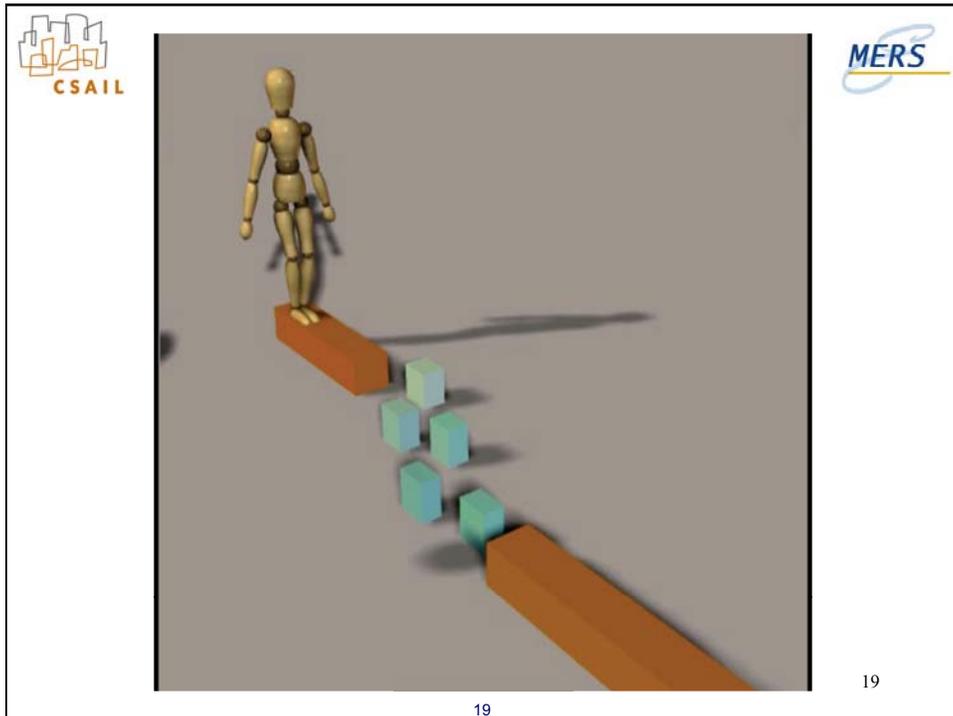
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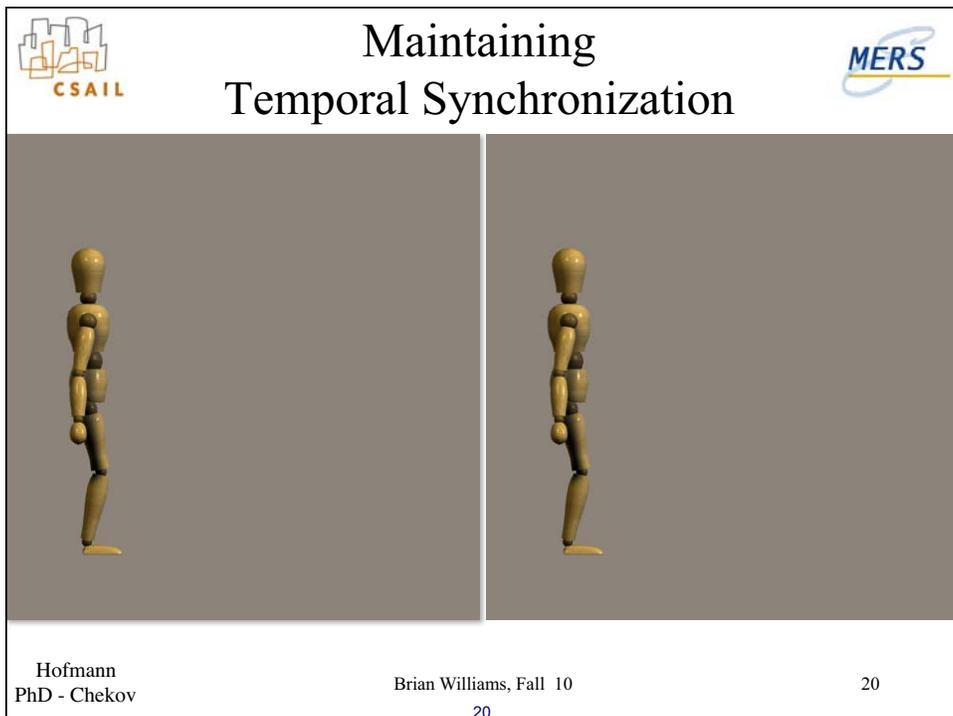
Courtesy JPL



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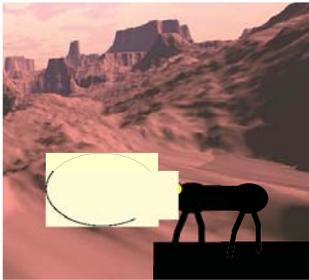
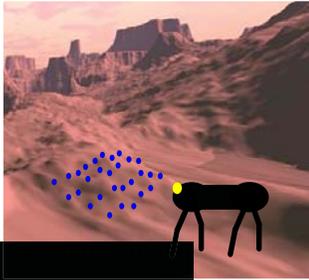


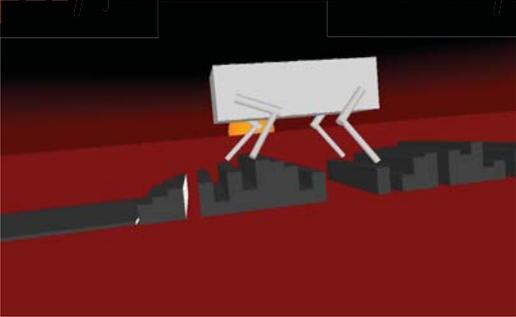
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CDIO Capstone: Moretta





Hofmann
PhD - Chekov

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3. Science Explorers

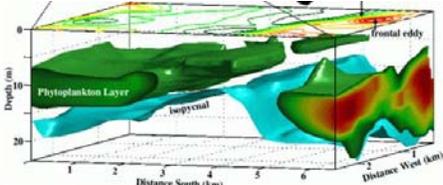


MBARI Dorado-class AUV:

- 6000m rated
- 20 hour operation
- Multibeam Sonars
- 3+ knots speed

Challenges:

- Long mission duration
- Limited communication
- GPS unavailable
- Uncertainty
 - tides and currents
 - estimation error



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Robust, Goal-directed Deep Sea Exploration



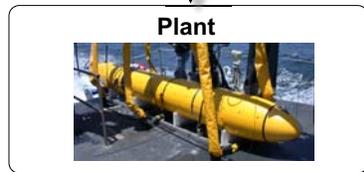
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00:00 Go to  $x_1, y_1$ 
00:20 Go to  $x_2, y_2$ 
00:40 Go to  $x_3, y_3$ 
...
04:10 Go to  $x_n, y_n$ 

```

Command script

Commands



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Leaute & Williams, AAAI 05

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Robust, Goal-directed Deep Sea Exploration



*“Remain in **mapping region** for at least 100s, then remain in **bloom region** for at least 50s, then return to **pickup region**. Avoid obstacles at all times”*

Qualitative State Plan

Model-based Executive

Observations

Commands

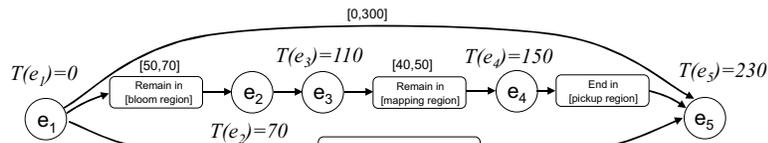
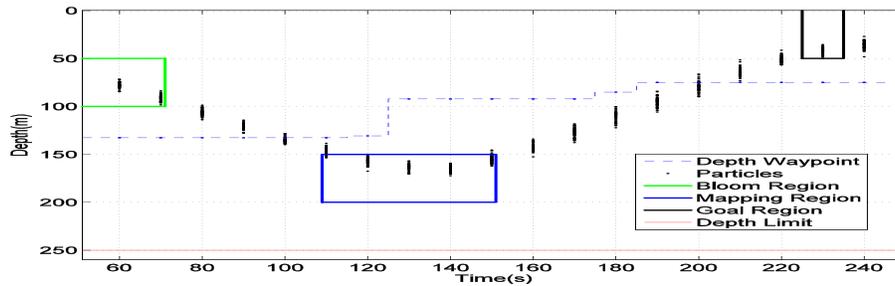


Optimal

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Blackmore PhD

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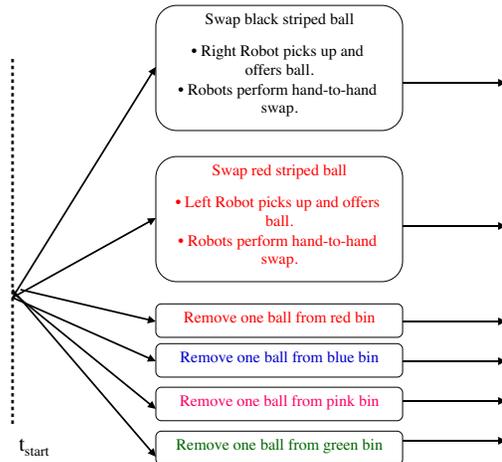
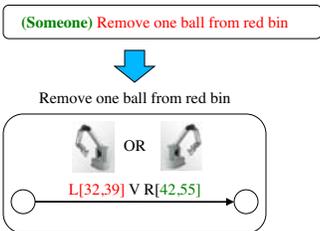
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Multi-Robot Teamwork



Agents choose and schedule activities

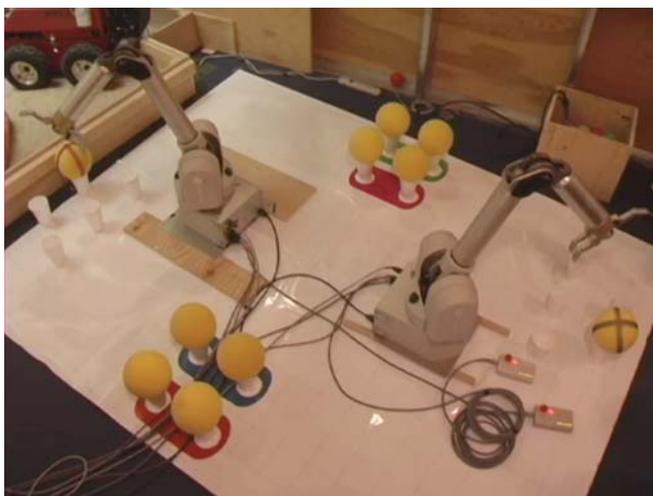


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Multi-Robot Teamwork



- Multi-modal
- Multi-agent
- Multi-task
- Multi-robot

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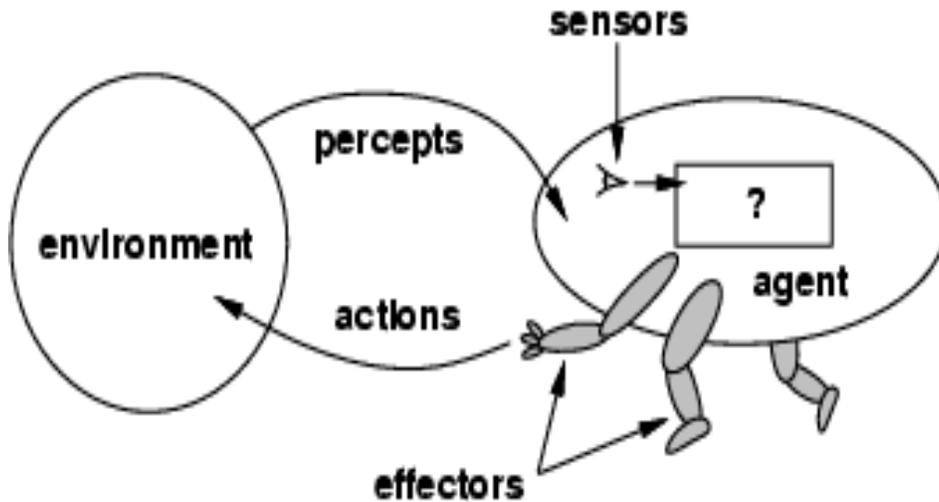


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Agent Paradigms

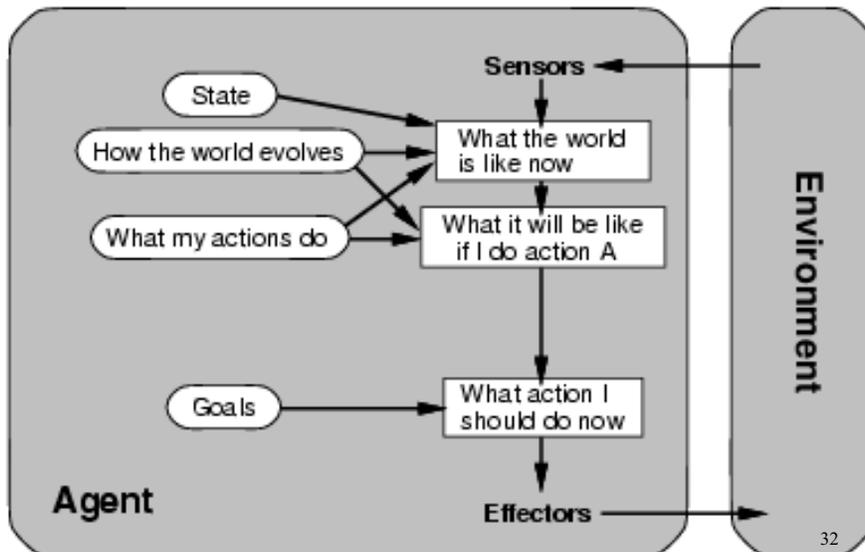


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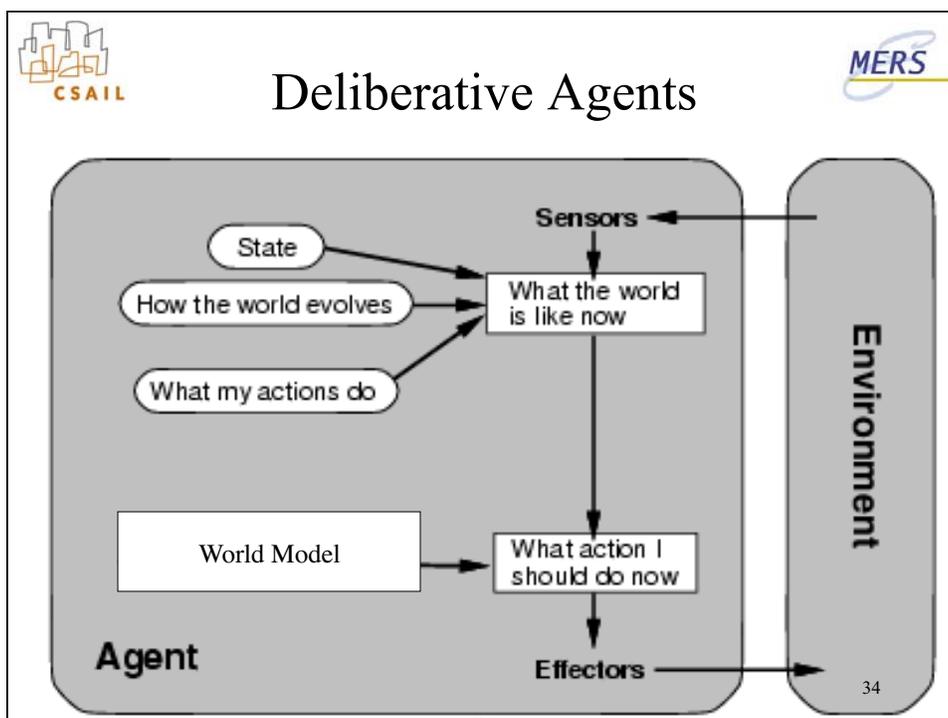
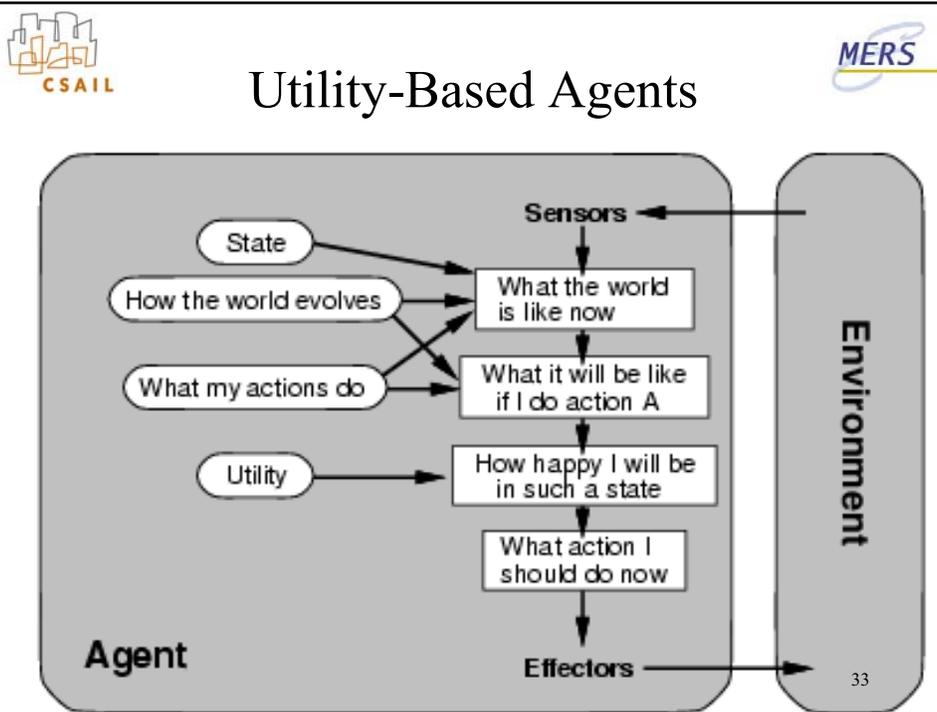
Goal-Oriented Agents



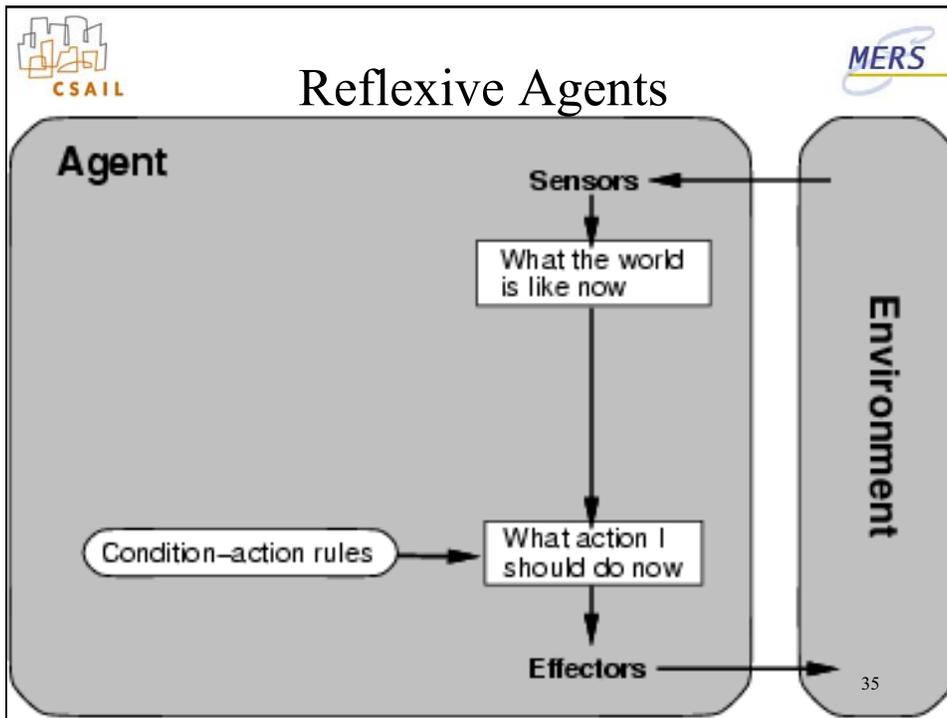
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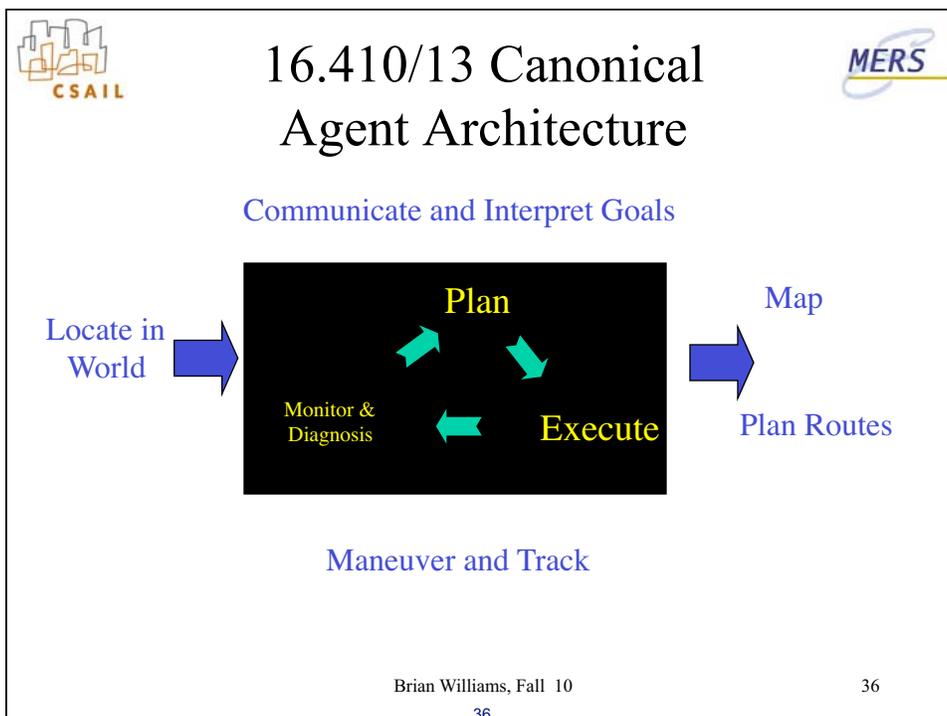
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Modeling Example: Robust, Goal-directed Deep Sea Exploration

“Remain in mapping region for at least 100s, then remain in bloom region for at least 50s, then return to pickup region. Avoid obstacles at all times”

Qualitative State Plan

Model-based Executive

Observations

Commands

Plant



Optimal

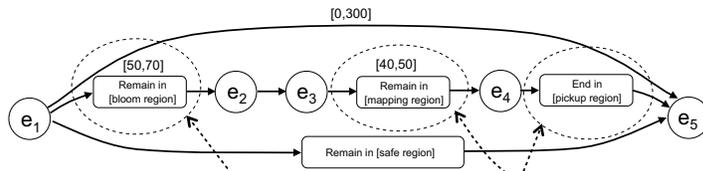
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Modeling Goal Behavior using Qualitative State Plans

A qualitative state plan is a model-based program that is unconditional, timed, and hybrid and provides flexibility in state and time.

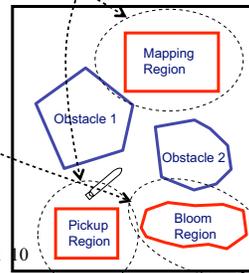


“Remain in bloom region for between 50 and 70 seconds. Afterwards, remain in mapping region for between 40s and 50s. End in the pickup region. Avoid obstacles at all times. Complete the mission within 300s”

Approach: Frame as Model-Predictive Control using Mixed Logic or Integer / Linear Programming.

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Modeling Goal-directed Planning as a Mathematical Program

$$\min_{\mathbf{U}} J(\mathbf{X}, \mathbf{U}) + H(x_T)$$

s.t. $J(\mathbf{X}, \mathbf{U})$ Cost function (e.g. fuel consumption)

Dynamics (Discrete time) $\forall_{0 \leq t \leq T-1} x_{t+1} = Ax_t + Bu_t$

Constraints $\bigwedge_{t=0}^T \bigwedge_{i=0}^N \bigvee_{j=0}^M h_t^{iT} x_t \leq g_t^{ij}$

Mixed Logic or Integer

$\mathbf{X} = [x_0 \cdots x_t]$ State vector (e.g. position of vehicle)

$\mathbf{U} = [u_0 \cdots u_{t-1}]$ Control inputs

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Specify Building Blocks using Declarative Programming



- Mathematical Programming
- Constraint Programming
- Logic Programming
- Agent Programming
 - Model-based programming
 - Timed concurrent constraint programming
 - Golog
 - Temporal logic programming

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Solve Declarative Programs using:



Search + Inference

Search: try taking the subway, or
try taking the bus.

Inference: It takes 35 minutes to get to MIT,
20 min subway + 15 min walking.

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