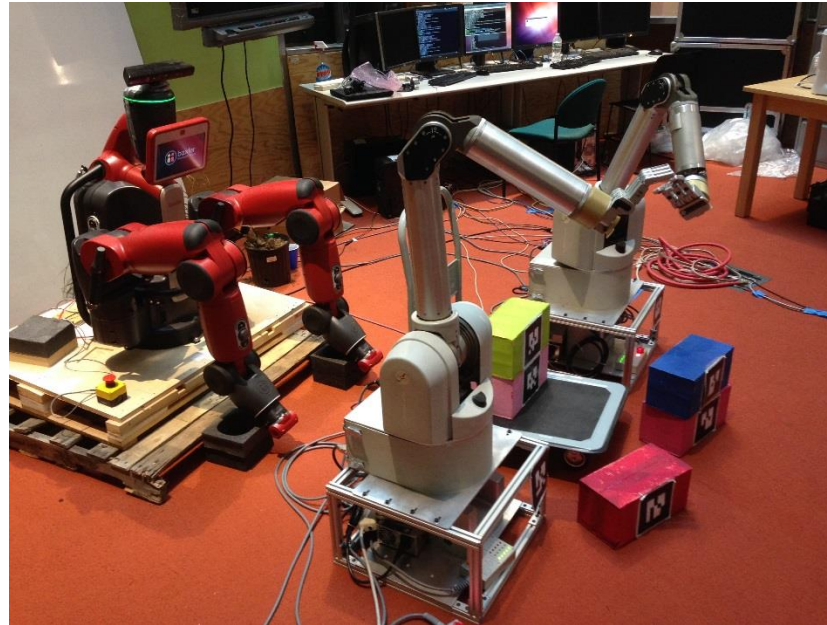


Cognitive Robotics

I.e., How to create “thinking” robots



Slide contributions:
Jacob Beal

Prof. Brian Williams and Steve Levine (TA)

Introduction

February 3rd, 2015.

Today's Assignments

Problems Sets:

- None! Problem Set #1 out next Monday, due in 1½ weeks .

Readings:

- **Today:** Williams, B. C., *et. al.*, “Model-based Programming of Fault-Aware Systems,” AI Magazine, 24, pp. 61-75, 2004.
- **Next:** Williams, B. C. and Ragno, R. “Conflict-directed A* and its Role in Model-based Embedded Systems,” Special Issue on Theory and Applications of Satisfiability Testing, Journal of Discrete Applied Math, 155, pp. 1562-1595, 2003.

Note:

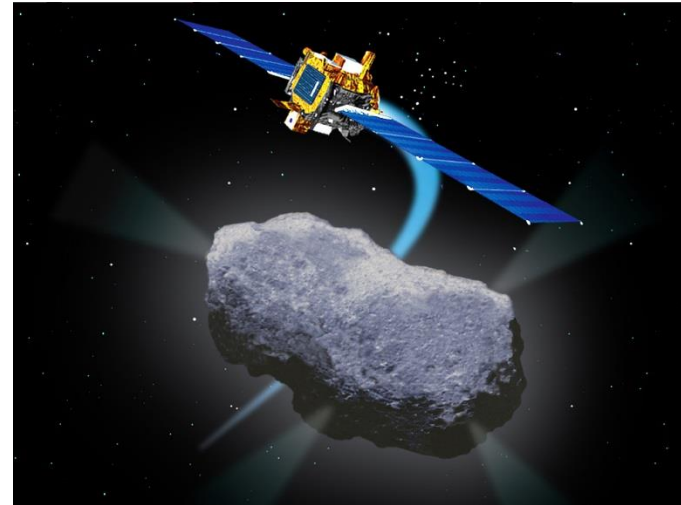
- Problem sets, readings and lecture slides posted on 16.412J course site.
- Background: open courseware, 16.410/13 Principles of Autonomy and Decision Making.

Outline

- **Course in a Nutshell**
- Course Structure and Logistics
- Programming Cognitive Robots
- Self-Adaptive and Self-Repairing Systems
- Programs that Monitor State

Coordinating Network Embedded Systems

- We are creating **vast networks** of embedded systems that perform **critical functions** over **long periods** of time.
- These long-lived systems achieve **robustness** by **coordinating a complex network** of devices.
- **Programming** these systems robustly is becoming an increasingly **daunting task**.



This image is in the public domain.



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

Plant



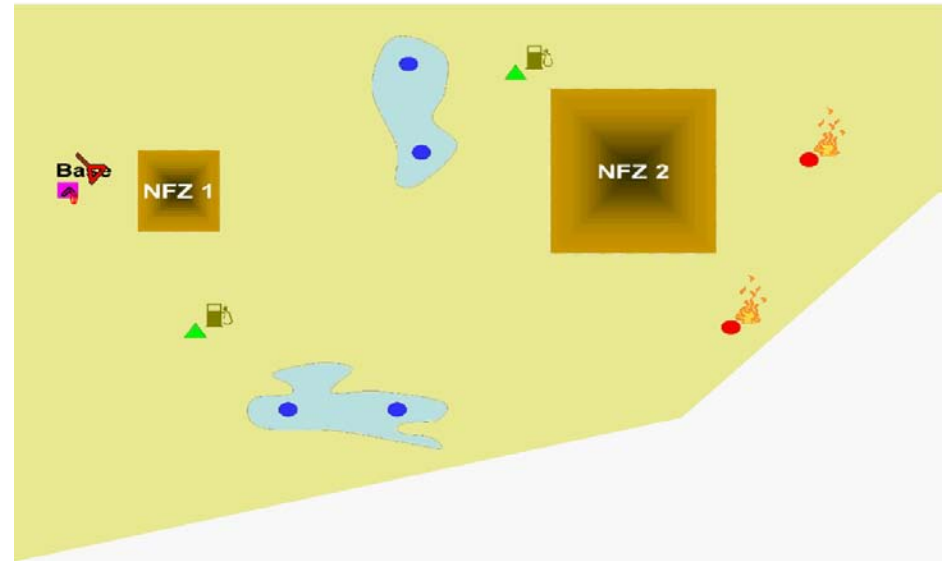
“Put out the fires at Locations 1 and 2, and return to the base in an hour. Avoid no fly zones. Here is a map, including reservoirs and gas stations.”
Qualitative State Plan

Model-based Executive

Observations

Commands

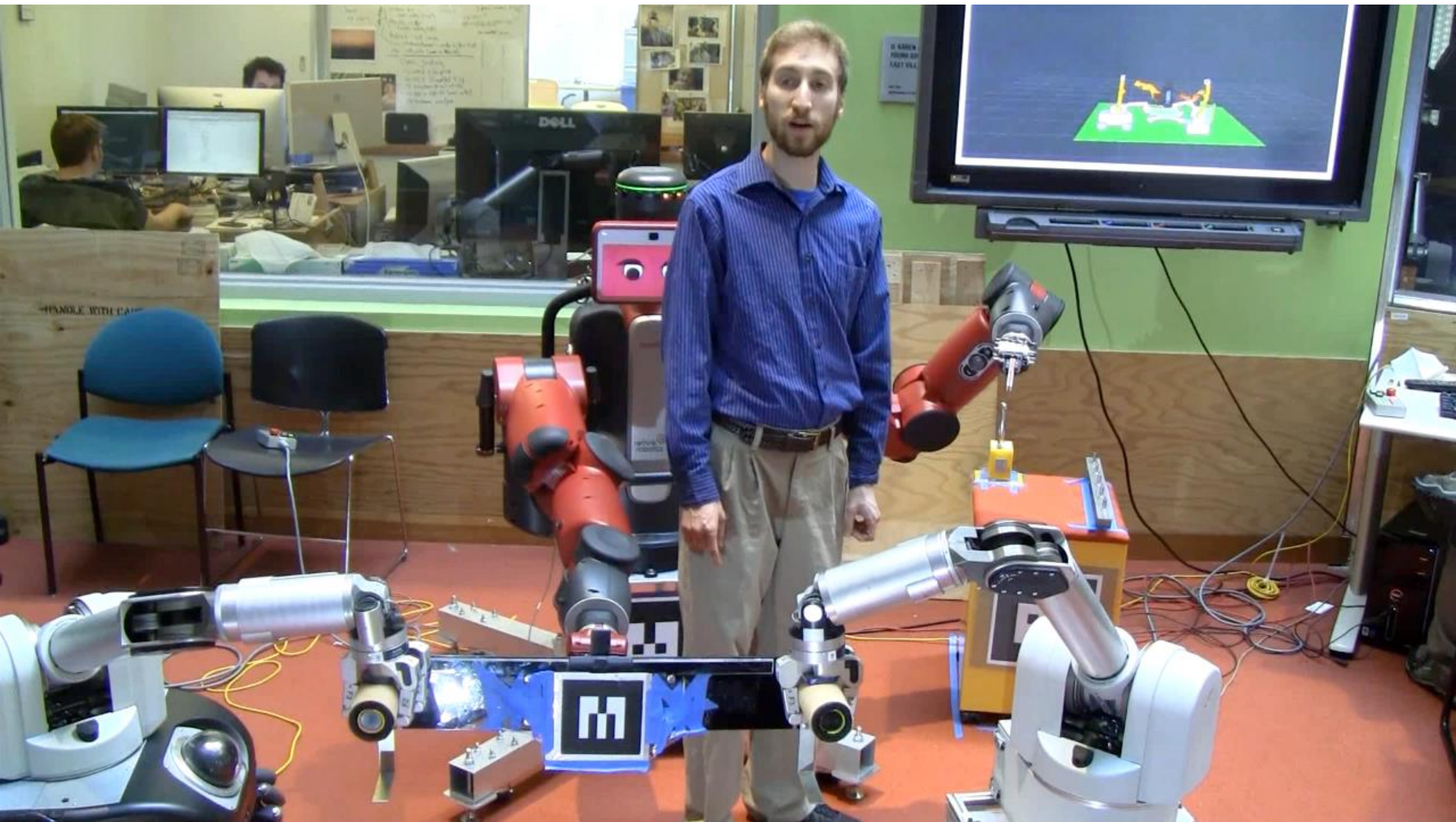
Plant



[Leaute & Williams, AAAI 05]

Robustness through Collaboration





Examples of Cognitive Robotic Systems

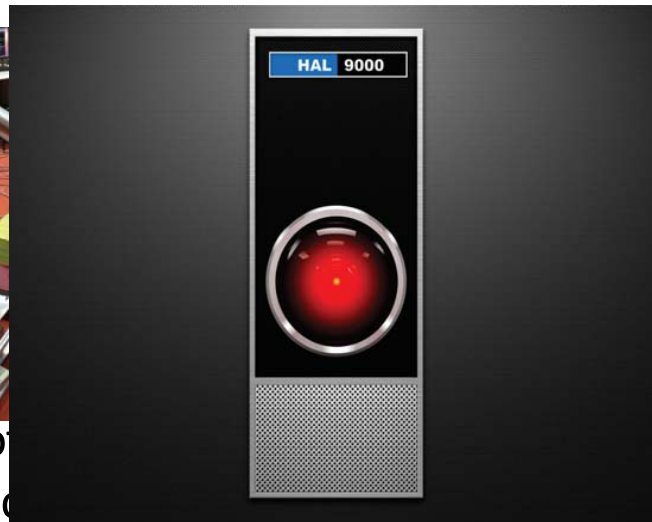


es:
irected
nt
nsitive
orative
formation

Information-gathering
Scouts



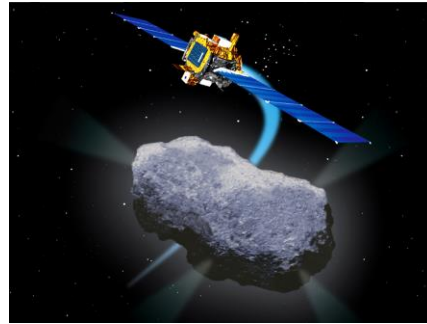
Human-Robot
For Manufact



Grids
Homes

© sources unknown. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.

Programming Cognitive Systems



This image is in the public domain.

1. Embedded **programming languages** elevated to the **goal-level** through **partial specification** and operations on **state** (RMPL).
2. Language **executives** that achieve robustness by reasoning over **constraint-based models** and by **bounding risk** (Enterprise).
3. Interfaces to support **human interaction fluidly** and at the **cognitive level** (Uhura, Pike ...).

Outline

- Course in a Nutshell
- **Course Structure and Logistics**
- Programming Cognitive Robots
- Self-Adaptive and Self-Repairing Systems
- Programs that Monitor State

Crash Course in Autonomy

I.e., Programming Cognitive Robots



Prof. Brian Williams & Eric Timmons

Erez Karpas, Andrew Wang, Peng Yu, Steve Levine, Pedro Santana, Simon Fang,
Enrique Gonzales, David Wang and Peng Yu.

Introduction

January 12, 2015.

About

- How to program cognitive robots in terms of goals, to perform complex tasks.
 - Intuitions underlying how robots “reason.”
 - ~~Exposure to basic computational concepts.~~
- 16:412J: State-of-the-art reasoning methods.

Driven by a Grand Challenge



© sources unknown. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.



© National Geographic Partners, LLC. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.

Embedded in simulations and hardware



Working out-of-the-box with your '15 holiday gift

© Parrot SA. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.

16:412J: Driven by a Grand Challenge



RobOrienteering

© British Orienteering. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.

16:412J: Embedded in simulations and hardware



Works with a lot of elbow grease and a bit of luck

Traditional Programming

TODAY

Programs that Monitor State

Programs with Flexible Time

Programs with Goal States

Programs with Continuous State

Programs that Collaborate

Advanced lectures...

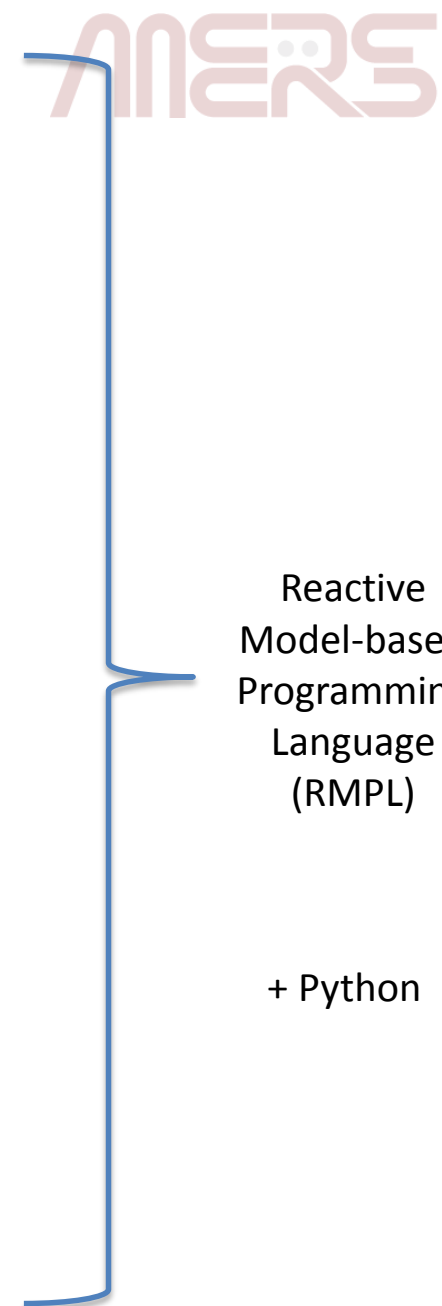
Risk-bounded Programming

“Cognitive” Programming

END OF SEMESTER

Grand Challenge

TIME



You'll walk out knowing how to

- program cognitive robots by specifying goals and models
- Research, describe, and implement decision-making algorithms that enable robots to monitor, plan and coordinate in the real world
- implement complex missions with real robot.

Be warned – bleeding edge!

A clipart of blooded knife removed due to copyright restrictions.

Policies, grading, and assignments

COURSE LOGISTICS

Websites & emails

- Stellar website
- Piazza (for course-related discussion)
- Staff email list

Lectures & Office Hours

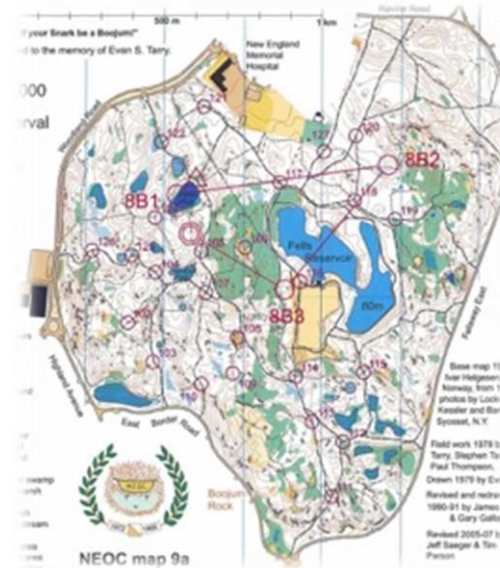
- **Students must attend lectures**
 - *Vital* to learning course material
 - Plan on attending all lectures
 - Expected to do assigned readings before lecture
- Randomly-scheduled 5-minute mini quizzes
 - Not difficult / stressful

Assignments: Problem sets

- Modeling exercises
- Using existing autonomy tools
- Implementing algorithms (Python)

Assignments: Grand Challenge

- Course culminates in grand challenge!
 - Orienteering theme
 - Simulation & real hardware



Assignments: Advanced Lectures

- What it's like doing research in autonomy
- Teams of 5-6 will:
 - Present full 80 minute lecture on researched topic
 - Implement topic (Python)
 - Release code, API, and tutorial / documentation for class
 - Will be used in grand challenge

Electronics use policy in lecture

Computers, tablets: *for note taking only!*

Please do not:

- Check email or facebook, surf web, watch adorable cat videos , etc.

Research shows it also **distracts others** nearby

Grading

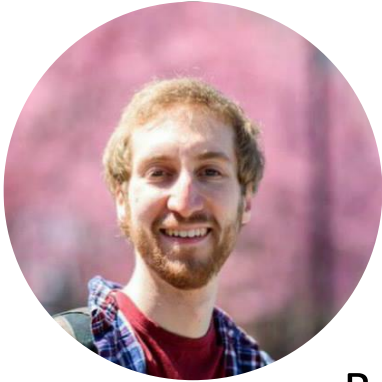
Item	Weight
Participation & attendance (mini quizzes)	10%
Problem sets	40%
Advanced lecture & implementation	30%
Grand challenge	20%

* Staff reserves the right to consider other factors & adjust formula

Collaboration Policy

- Collaboration allowed, such that you:
 - Acknowledge collaborators
 - Involved in all aspects of work (no dividing up)
 - Write your own solutions
- Advanced lectures & grand challenge
 - Working in teams, expect equal contributions
- Course bibles prohibited

TA: Steve Levine



4th year Ph.D student in CS (MERS group)

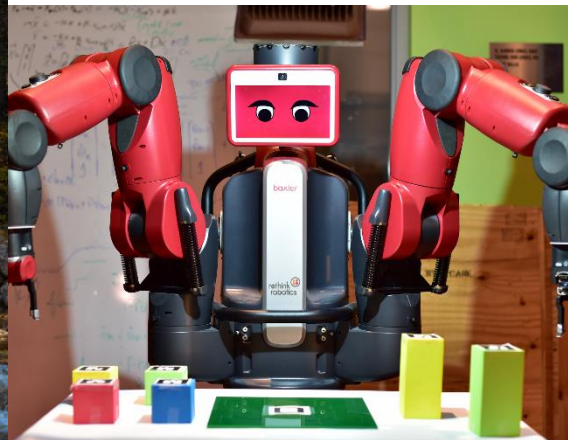
Research:

Intent recognition & adaptation for robots

B.S. from MIT in '11, course 6

M.Eng from MIT in '12, course 6

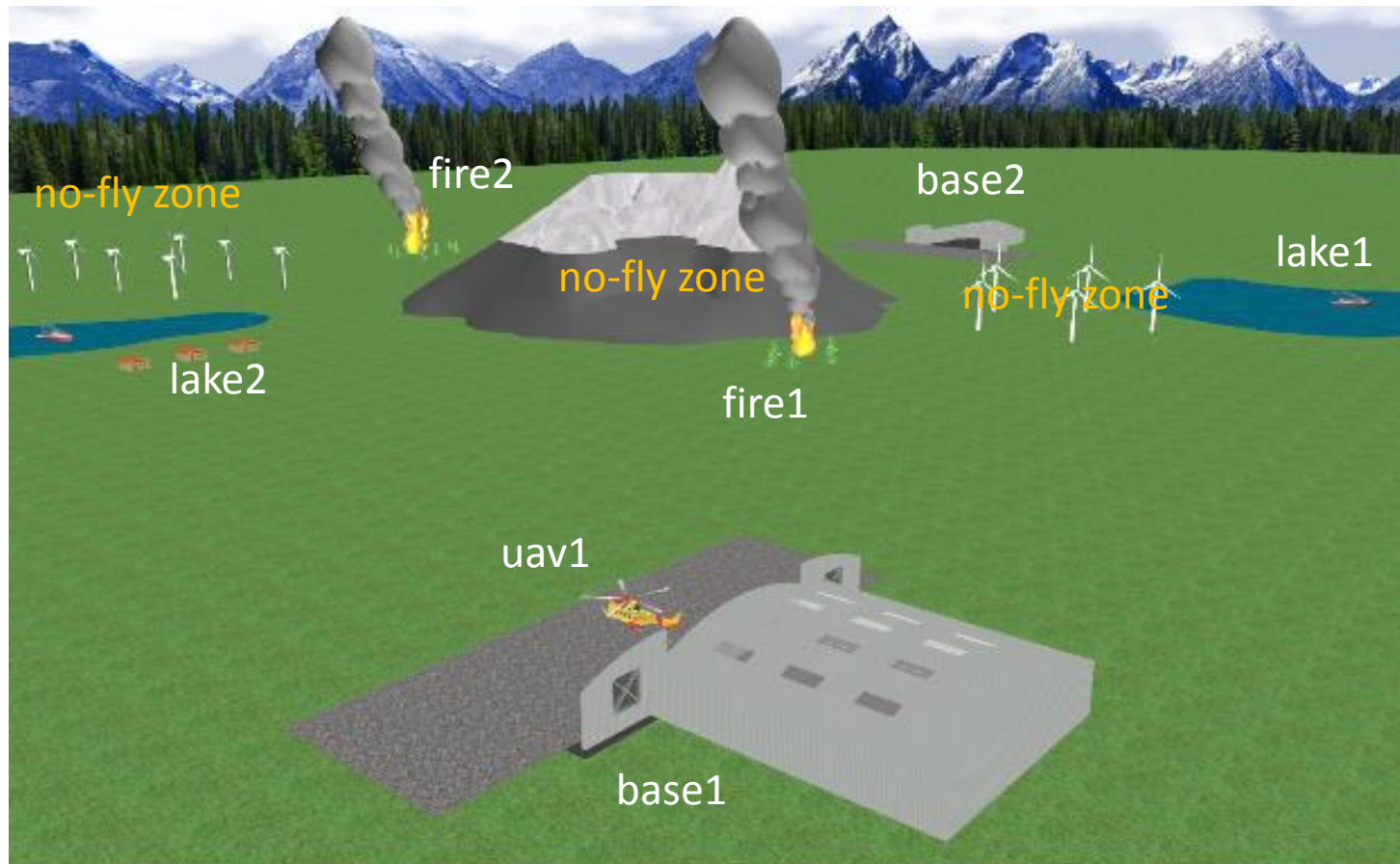
9th year (!) at MIT! So old!



Outline

- Course in a Nutshell
- Logistics
- **Programming Cognitive Robots**
- Self-Adaptive and Self-Repairing Systems
- Programs that Monitor State

Programs on State - Firefighting Scenario



Firefighting in RMPL:

Traditional Imperative program



```
class Main{
    UAV uav1;
    Lake lake1;
    Lake lake2;
    Fire fire1;
    Fire fire2;
    ...

    method run() {
        sequence{
            uav1.takeoff();
            uav1.fly_base1_to_lake1();
            uav1.load_water(lake1);
            uav1.fly_lake1_to_fire1();
            uav1.drop_water_high_altitude(fire1);
            uav1.fly_fire1_to_lake1();
            uav1.load_water(lake1);
            uav1.fly_lake1_to_fire1();
            uav1.drop_water_low_altitude(fire1);

            uav1.fly_fire1_to_lake2();
            uav1.load_water(lake2);
            uav1.fly_lake2_to_fire2();
            uav1.drop_water_high_altitude(fire2);
            uav1.fly_fire2_to_lake2();
            uav1.load_water(lake2);
            uav1.fly_lake2_to_fire2();
            uav1.drop_water_low_altitude(fire2);
            uav1.fly_fire2_to_base1();
            uav1.land();
        }
    }
}
```

Firefighting in RMPL: A Program on State

```
class Main{
  UAV uav1;
  Lake lake1;
  Lake lake2;
  Fire fire1;
  Fire fire2;
  ...

  method run() {
    sequence{
      (fire1 == out);
      (fire2 == out);
      (uav1.flying == no &&
       uav1.location == base_1_location);
    }
  }
}
```

Firefighting in RMPL: Setup & Initial Conditions

```

class Main{
  UAV uav1;
  Lake lake1;
  Lake lake2;
  Fire fire1;
  Fire fire2;
  ...

  method run() {
    sequence{
      (fire1 == out);
      (fire2 == out);
      (uav1.flying == no &&
       uav1.location == base_1_location);
    }
  }
}

```

```

Main (){
  uav1 = new UAV();
  uav1.location= base_1_location;
  uav1.flying = no;
  uav1.loaded = no;

  lake1 = new Lake();
  lake1.location = lake_1_location;

  lake2 = new Lake();
  lake2.location = lake_2_location;

  fire1 = new Fire();
  fire1.location = fire_1_location;
  fire1 = high;

  fire2 = new Fire();
  fire2.location = fire_2_location;
  fire2 = high;
}

```

Firefighting in RMPL: Model of Actions

```

class UAV {
  Roadmap location;
  Boolean flying;
  Boolean loaded;

  primitive method takeoff()
    flying == no => flying == yes;

  primitive method land()
    flying == yes => flying == no;

  primitive method load_water(Lake lakespot)
    ((flying == yes) && (loaded == no) && (lakespot.location == location)) => loaded == yes;

  primitive method drop_water_high_altitude(Fire firespot)
    ((flying == yes) && (loaded == yes) && (firespot.location == location) && (firespot == high))
    => ((loaded == no) && (firespot == medium));

  primitive method drop_water_low_altitude(Fire firespot)
    ((flying == yes) && (loaded == yes) && (firespot.location == location) && (firespot == medium))
    => ((loaded == no) && (firespot == out));

  #MOTION_PRIMITIVES(location, fly, flying==yes)
}

```

```

class Lake {
  Roadmap location;
}

class Fire{
  initial value high;
  value medium;
  value out;
  Roadmap location;
}

```

Firefighting in RMPL: Model of Actions

```

class UAV {
  Roadmap location;
  Boolean flying;
  Boolean loaded;

  primitive method takeoff()
    flying == no => flying == yes;

  primitive method land()
    flying == yes => flying == no;

  primitive method load_water(Lake lakespot)
    ((flying == yes) && (loaded == no) && (lakespot.location == location)) => loaded == yes;

  primitive method drop_water_high_altitude(Fire firespot)
    ((flying == yes) && (loaded == yes) && (firespot.location == location) && (firespot == high))
    => ((loaded == no) && (firespot == medium));

  primitive method drop_water_low_altitude(Fire firespot)
    ((flying == yes) && (loaded == yes) && (firespot.location == location) && (firespot == medium))
    => ((loaded == no) && (firespot == out));

  #MOTION_PRIMITIVES(location, fly, flying==yes)
}

```

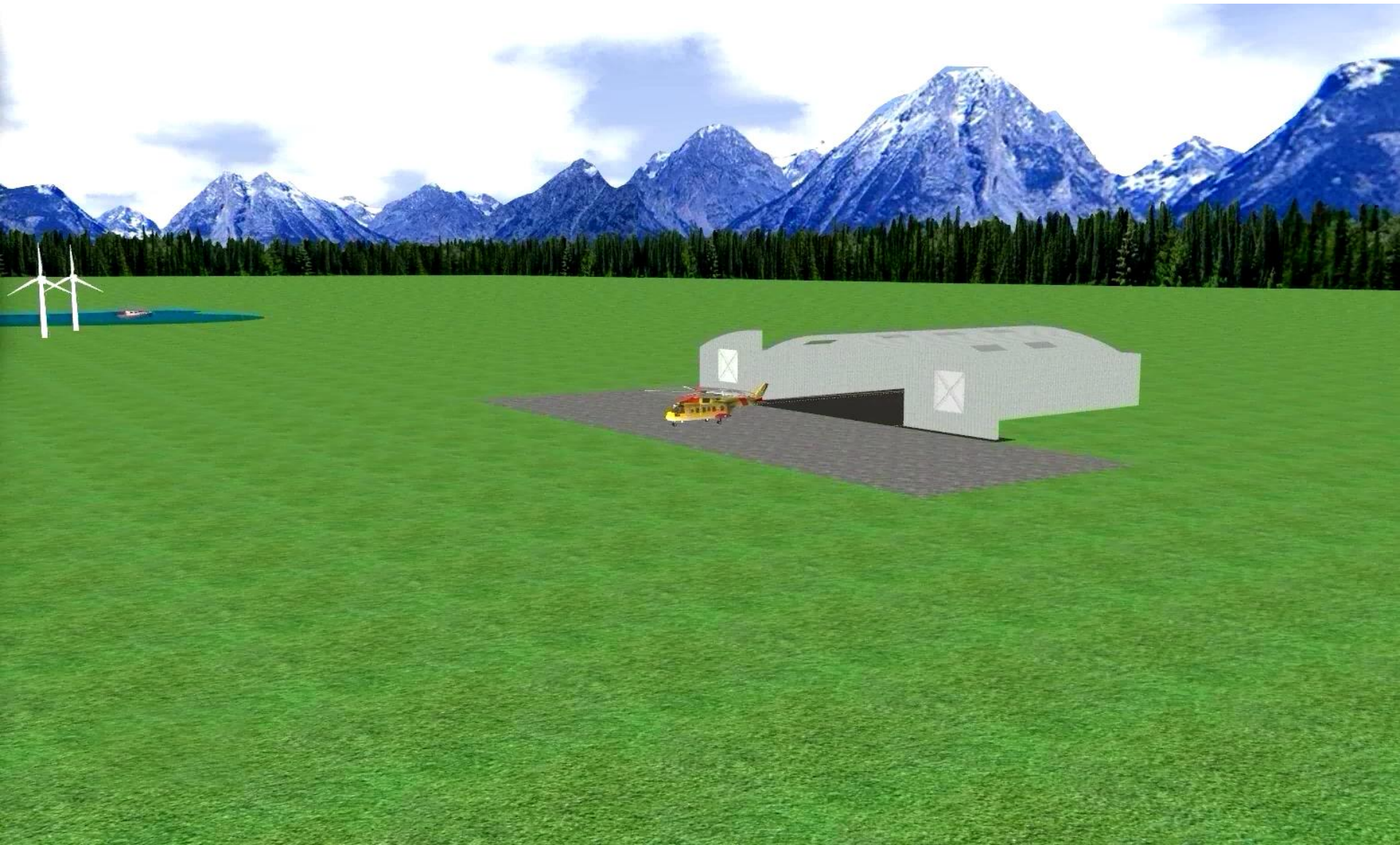
```

class Lake {
  Roadmap location;
}

class Fire{
  initial value high;
  value medium;
  value out;
  Roadmap location;
}

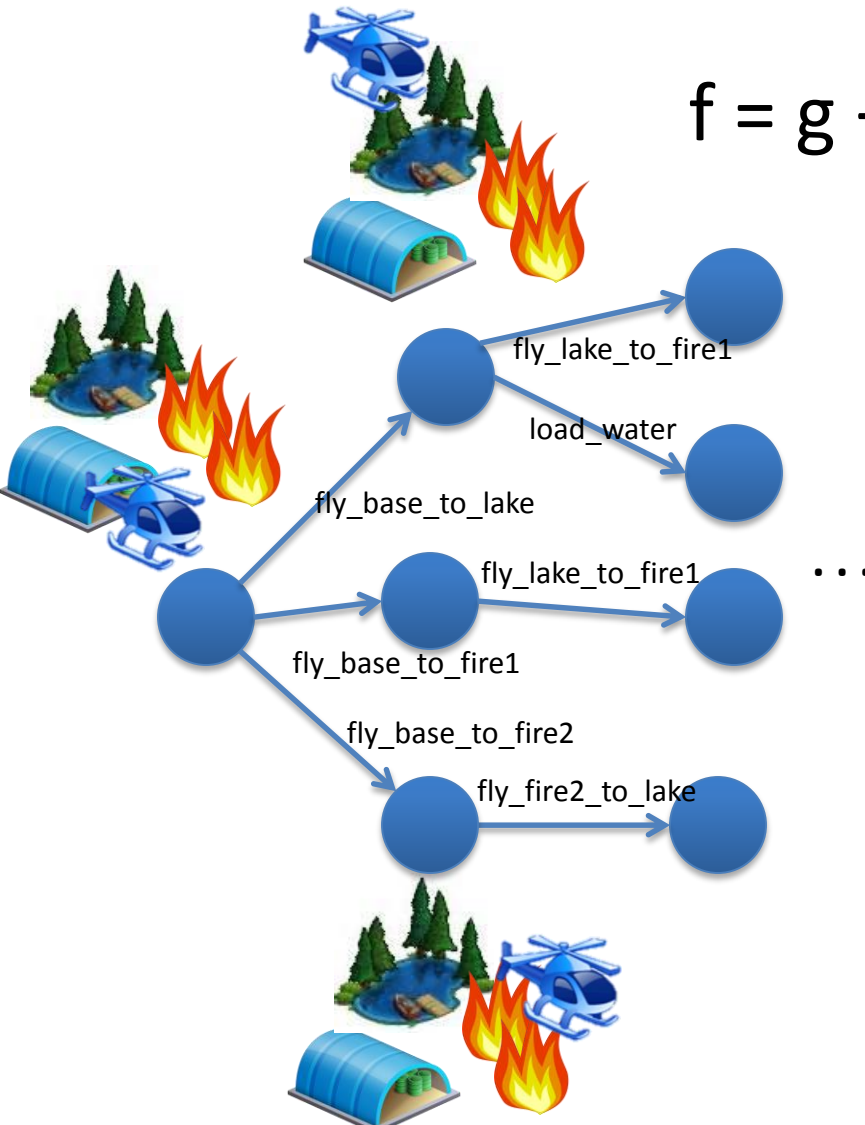
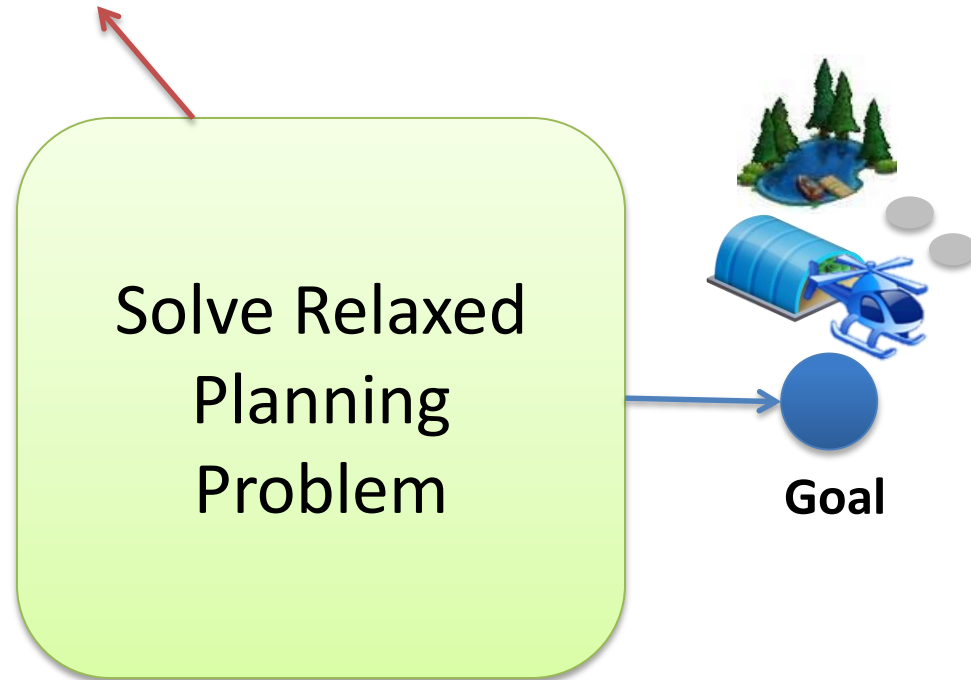
```

Simulation Testing: Firefighting Scenario



Decision-making algorithm: Activity Planning

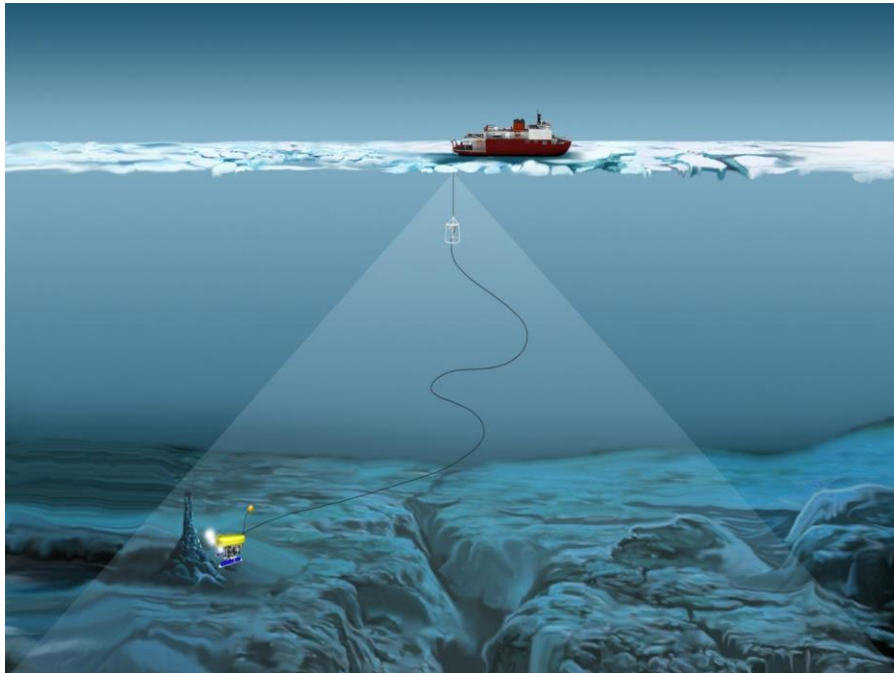
$$f = g + h$$



Outline

- Course in a Nutshell
- Logistics
- Programming Cognitive Robots
- **Self-Adaptive and Self-Repairing Systems**
- Programs that Monitor State

“Autonomous” vehicles explore far away places .. but often end in disaster!



© Woods Hole Oceanographic Institution. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.

What you should *really* learn from immune systems about adversarial design

Jacob Beal

NDIST
December, 2015

Courtesy of Jacob Beal. Used with permission.

Raytheon
BBN Technologies

Not for release: images shamelessly ganked from web

The Immune System

(as noticed by computer scientists)

Immune system cells

Complements
of Jake Beal

INSUFFICIENT



(Source: the Human Immune Response System www.uta.edu/chagas/images/immunSys.jpg)

But there's more to the immune system...

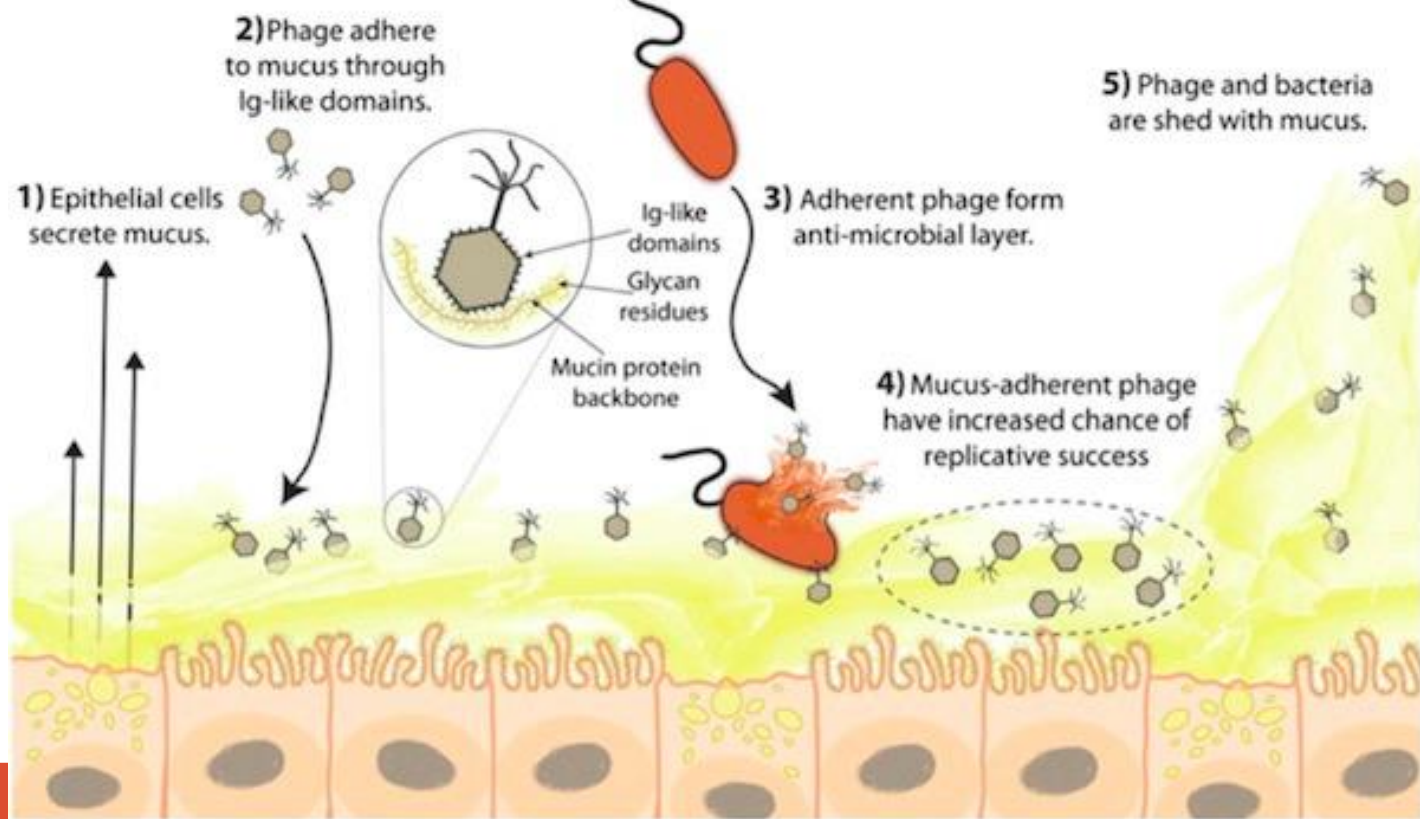


Complements
of Jake Beal

- Physical barriers
- Inhospitable environments
- Tolerance
- Death

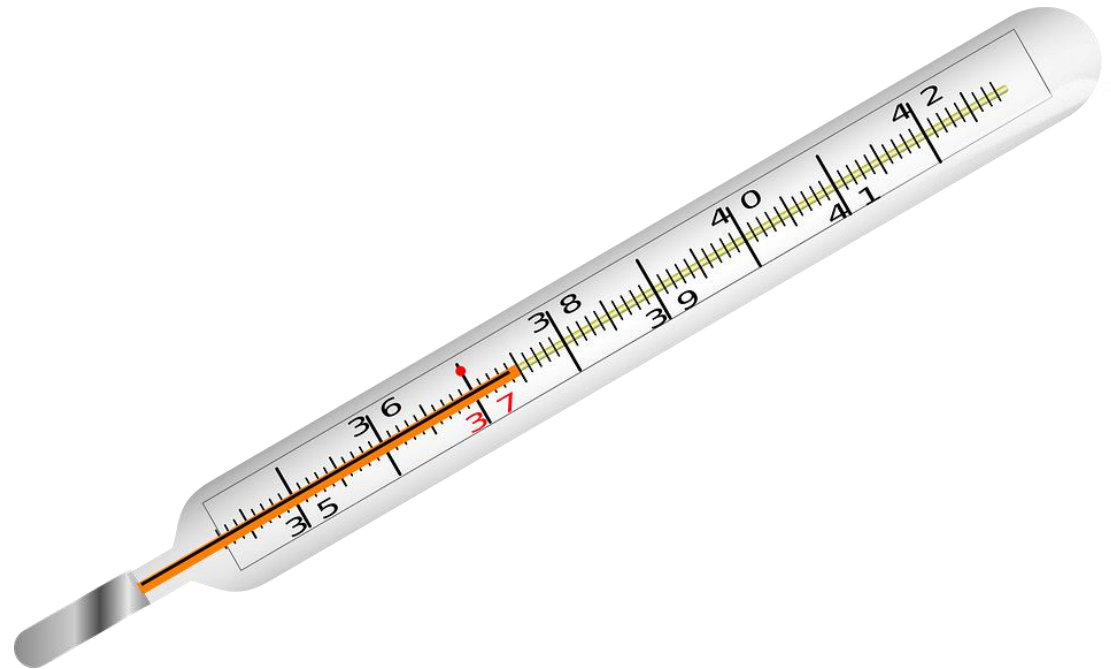
Physical Barriers

- Skin, gastrointestinal wall, blood/brain barrier
- Saliva, tears, mucus (nose, lungs, gut, . . .)
- Defense by discarding



Inhospitable environments

- Fever: high temperature inhibits bacterial growth
- Enzymes in saliva, tears, nasal secretions, perspiration, milk, sperm
- ...
- Skin is acidic



Tolerance

- Follicular mites
- Parasitic worms
- Dead viruses in the genome
- Microbiome
 - Gut commensals
 - Flora for various cavities

>3000 cryptic drug-like molecule gene clusters!

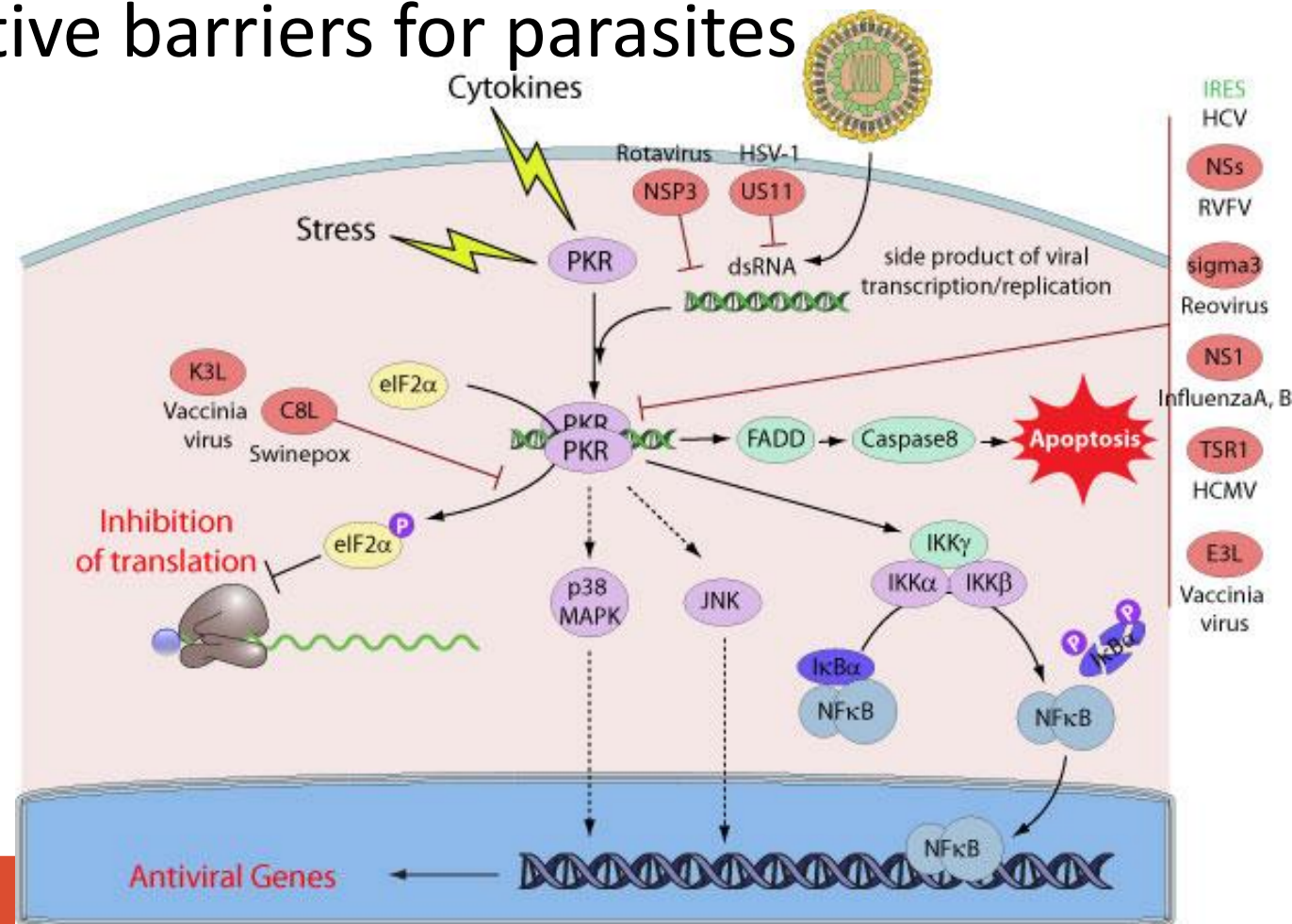
(Fischbach, Science, 2015)



Human follicular mite

Death

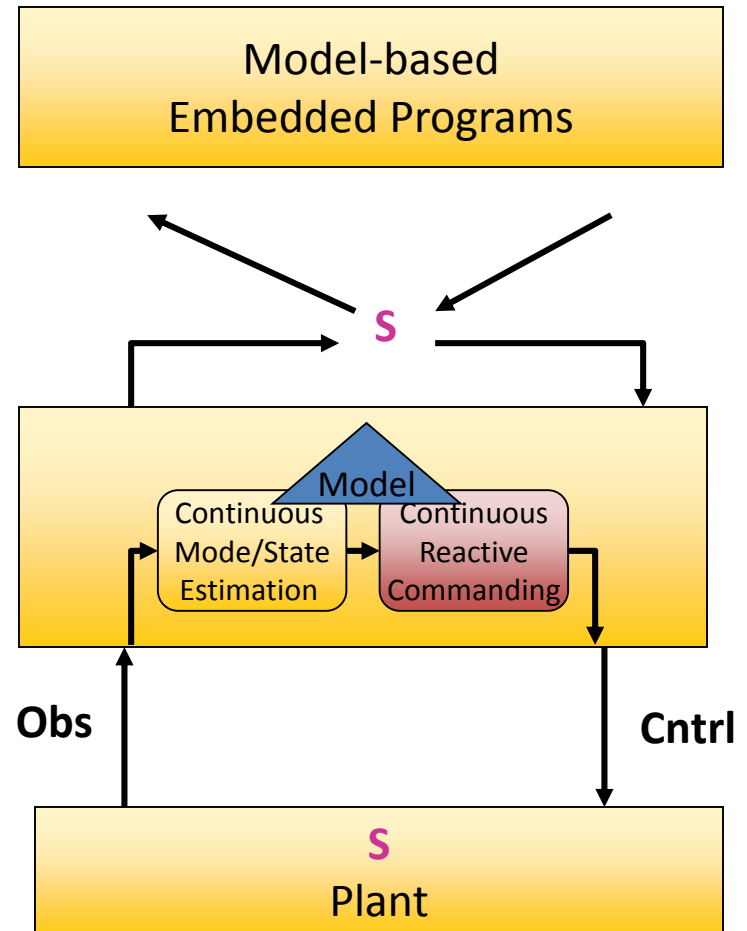
- Cell self-inhibition, suicide
- Reproductive barriers for parasites



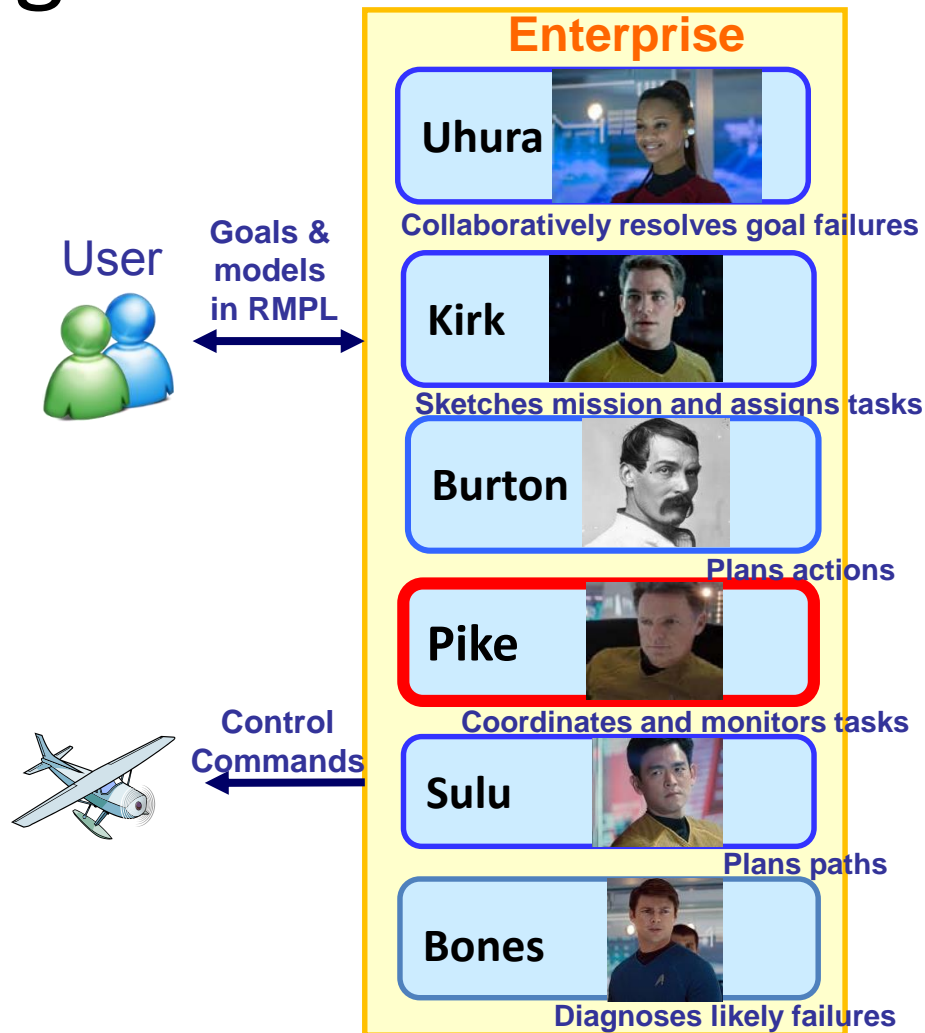
Model-based Programs Offer Layers of Defense

Languages that achieve robustness through decision layers that are:

- **Suspicious**
 - Monitor states and goals.
- **Adapt to disturbance**
 - Adjust timing
 - Select contingencies
- **State Aware**
 - Plan to achieve goals states.
- **Precise**
 - Achieve continuous goal states.
- **Collaborate**
 - Executes programs, revise goals and plan with humans.
- **Manage Risk**
 - Executes programs in uncertain environments with bounded risk.



A single “cognitive system” language and executive.



© source unknown. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.

Outline

- Course in a Nutshell
- Logistics
- Programming Cognitive Robots
- Self-Adaptive and Self-Repairing Systems
- **Programs that Monitor State**

When things go wrong...



Execution monitoring: detecting problems

- Something unexpected things happen
 - Not modeled in plan!
 - How to react?
- **Execution monitoring:** *detecting* when things go wrong
- **Replanning:** fixing the problem (later in this course)

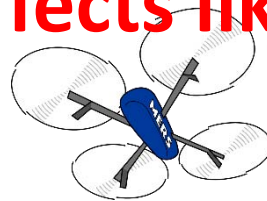


Volcano eruption

```

method run() {
  sequence {
    uav.launch();
    uav.fly_to_base_station();
    uav.pick_up_med_kit();
    uav.fly_to_hikers();
    uav.drop_off_med_kit();
  }
}

```



Actions have preconditions & effects like before



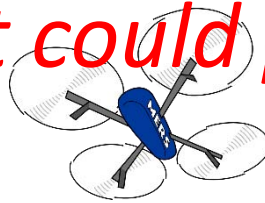
Program sequence of actions in RMPL

Volcano eruption



Base Station

What could possibly go wrong??



Launch Q

Q fly to Base Station

Q pick up med kit

Q fly to hikers

Q give med kit to climbers

Preconditions & effects of actions



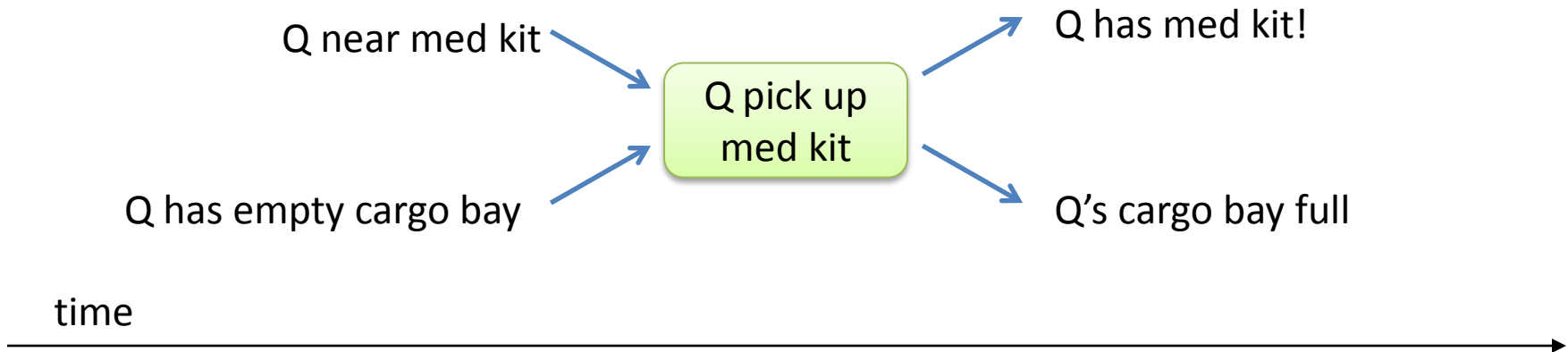
Preconditions & effects of actions

Q pick up
med kit

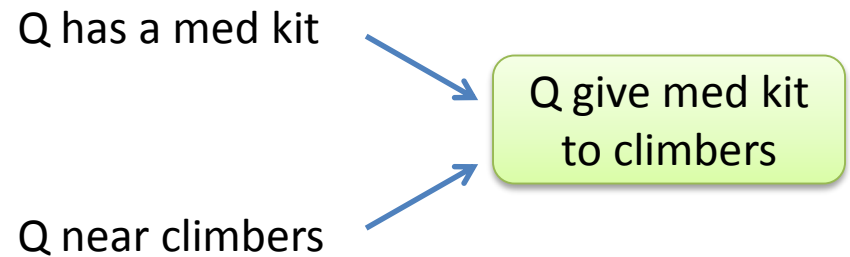
time



Preconditions & effects of actions



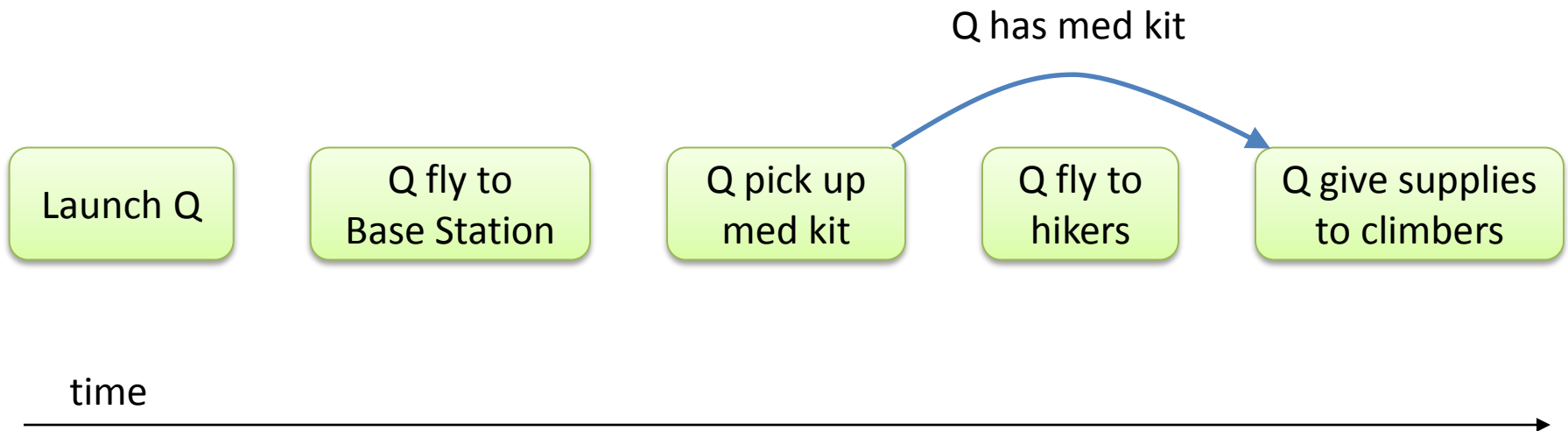
Preconditions & effects of actions



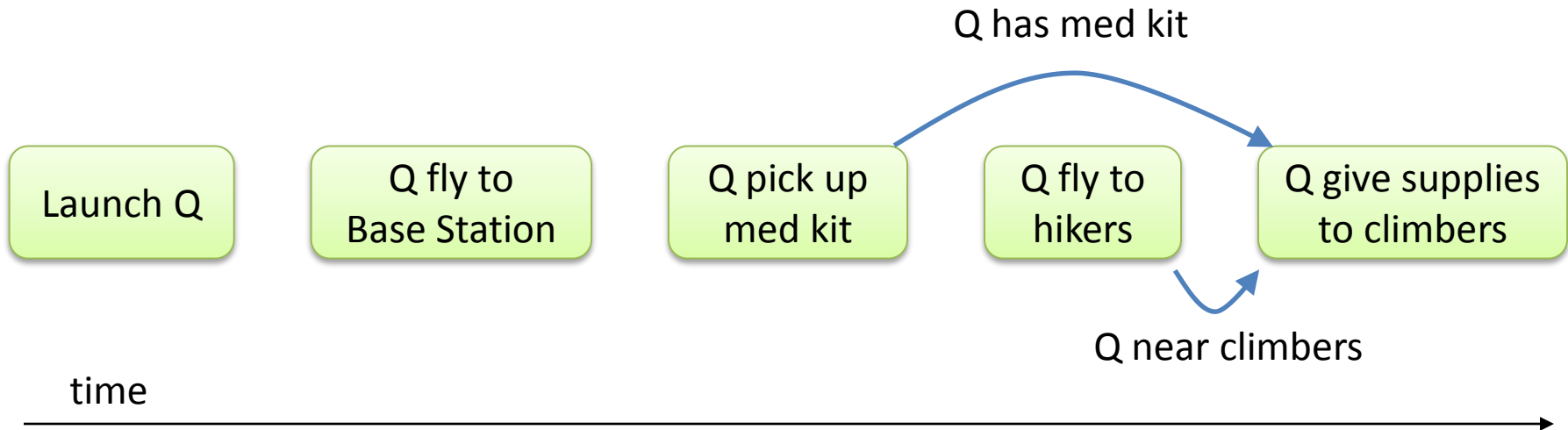
time



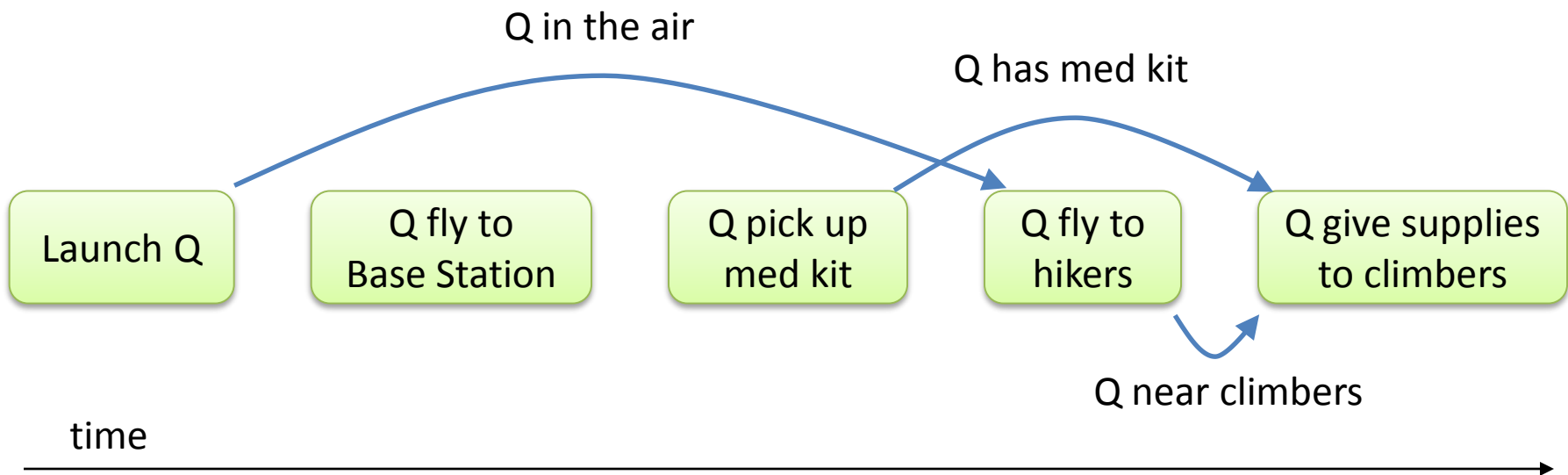
Preconditions & effects of actions



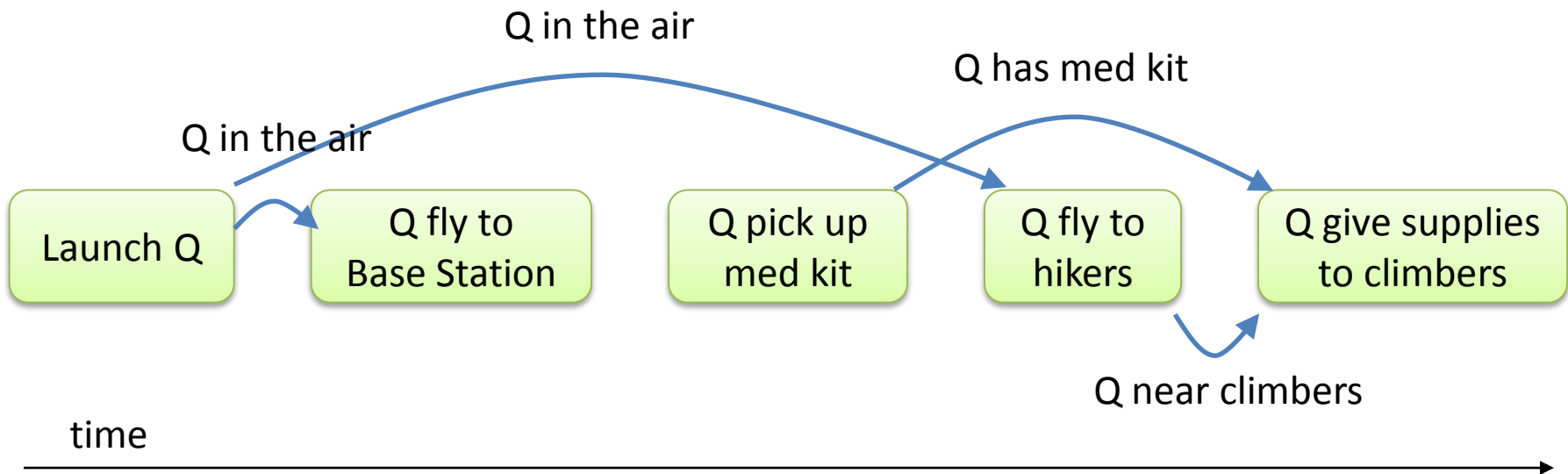
Where do preconditions come from?



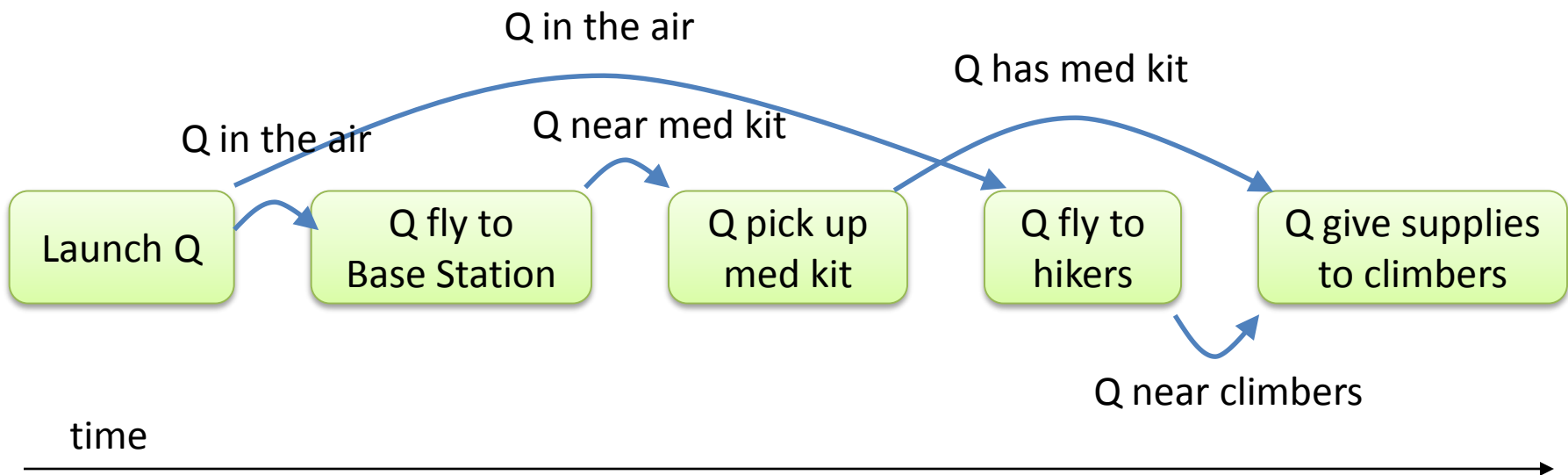
Where do preconditions come from?



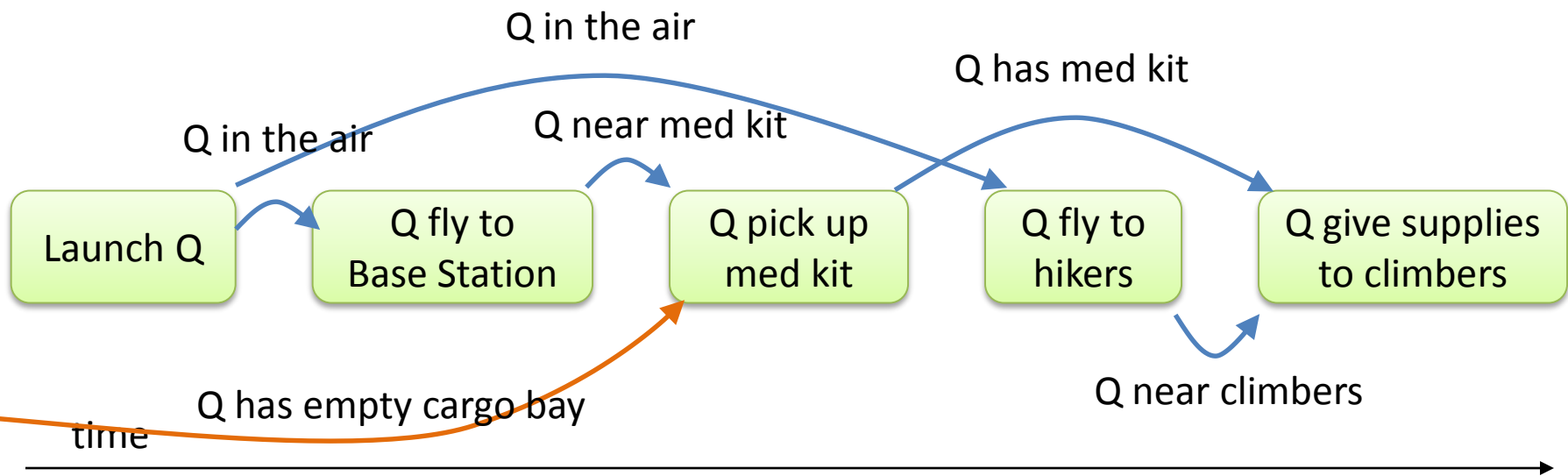
Where do preconditions come from?



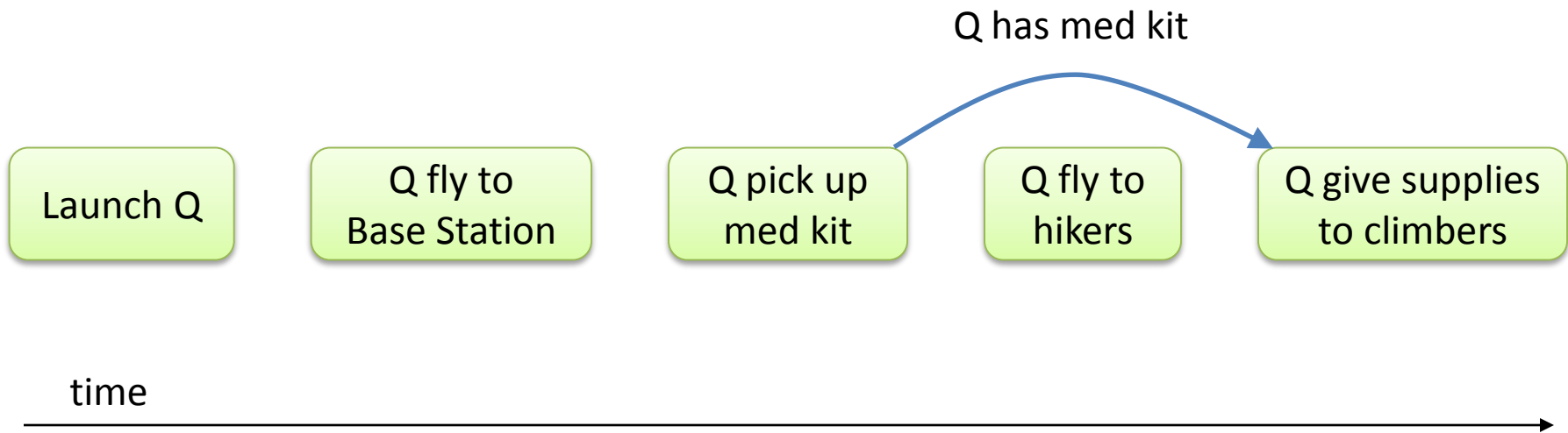
Where do preconditions come from?



Where do preconditions come from?

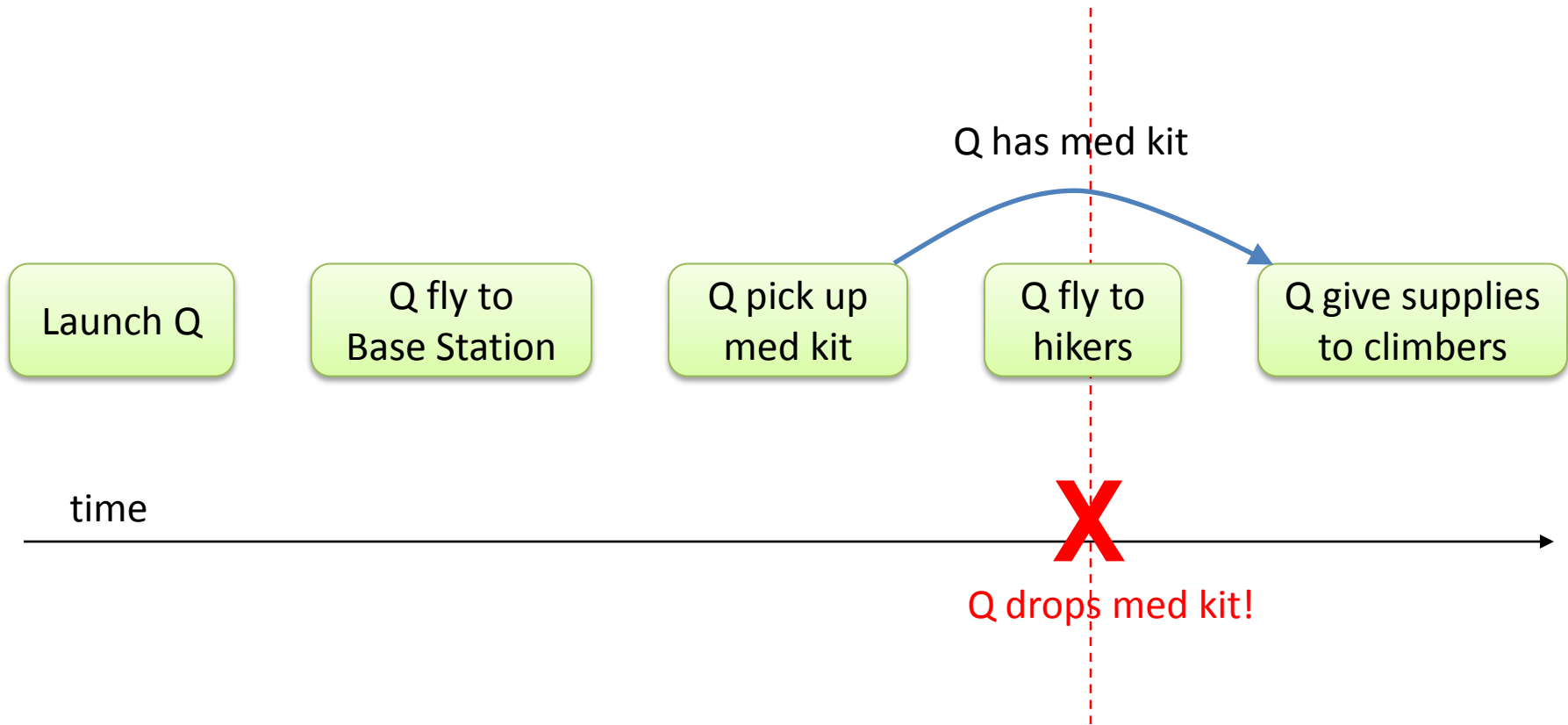


These are called “causal links”

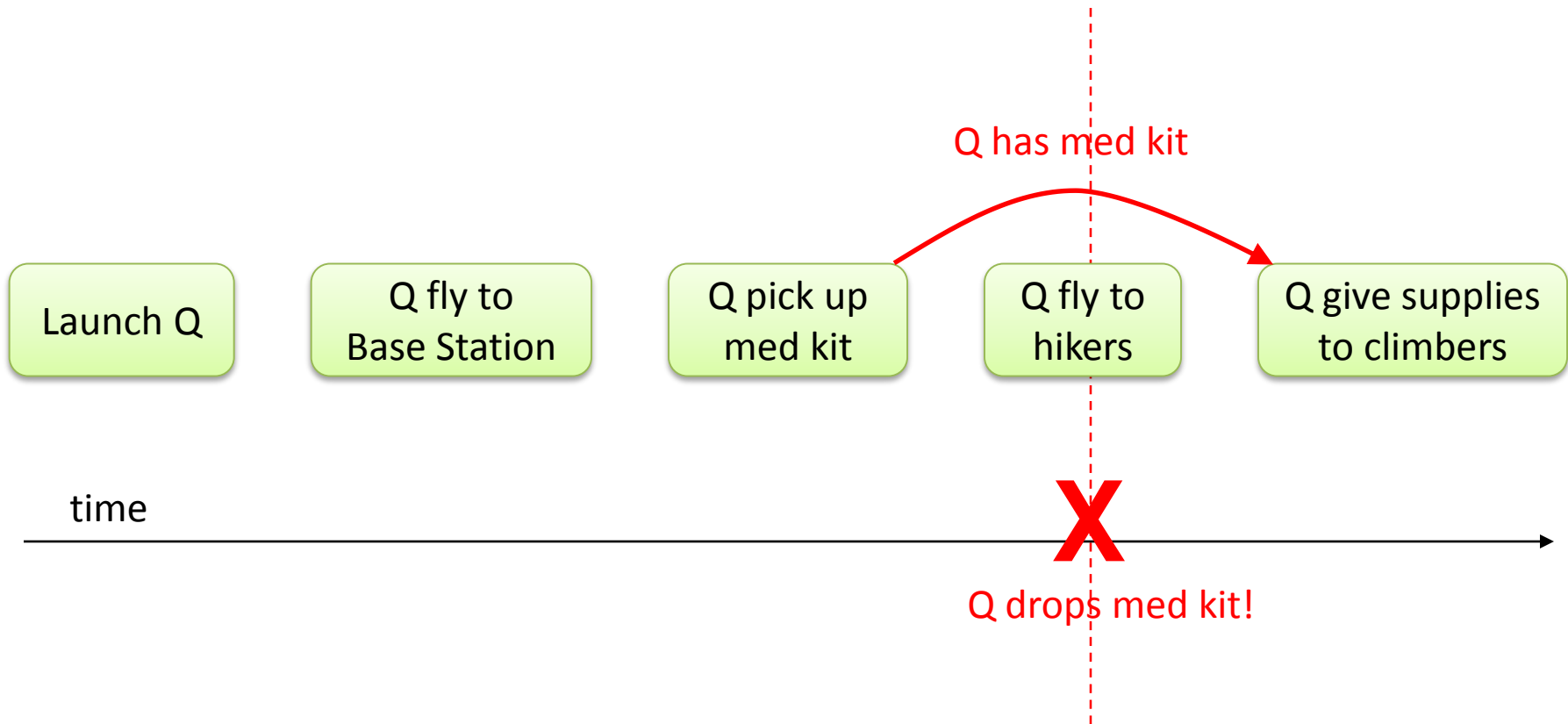


Causal link: one action produces something needed by a later action

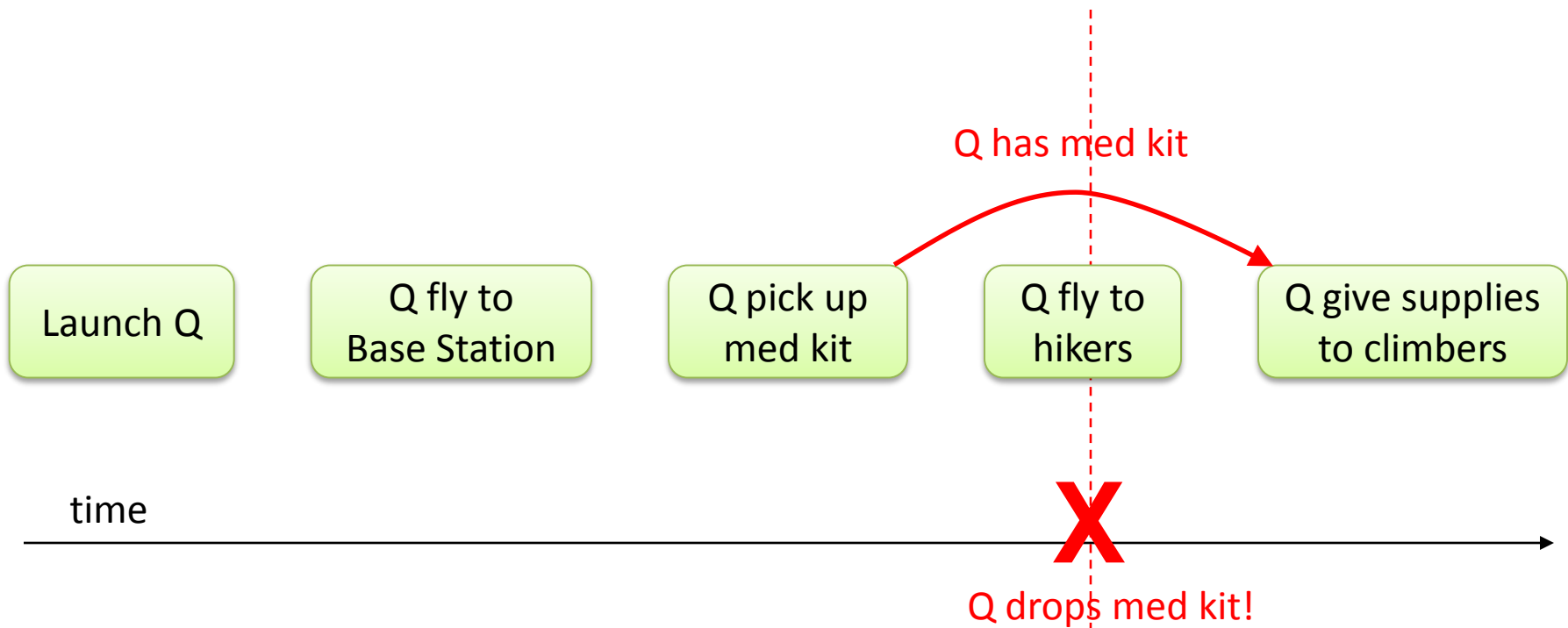
What happens if something goes wrong?



What happens if something goes wrong?



Causal links allow execution monitoring



If condition is violated during causal link, signal failure.

Causal links for execution monitoring

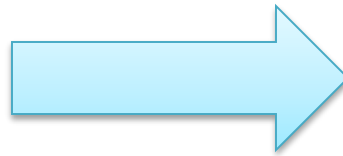
- Causal links tell you what needs to hold, and when
- Tells you what's *relevant* to the plan
- Can be used offline, for error checking

How do we do execution monitoring?

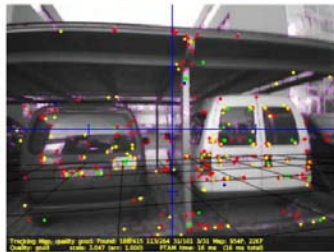
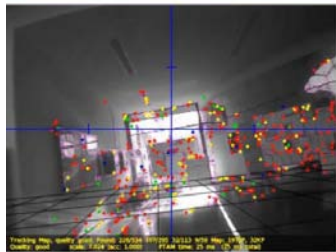
We need:

1. Sensing.
2. Action model.
3. Plan.

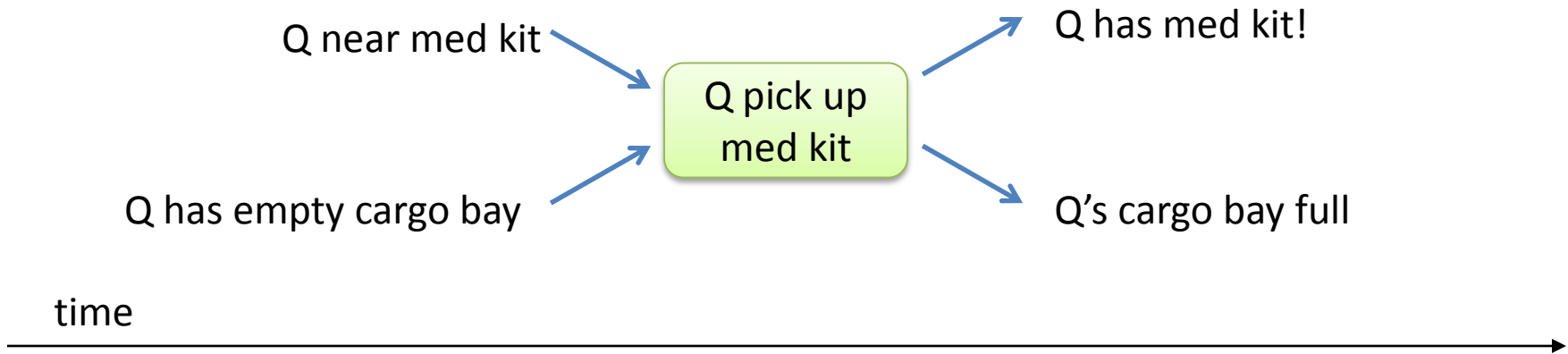
1. Sensing



“Q has medical kit”



2. Action model



3. Plan



How do we do execution monitoring?

Offline:

- Use **plan & action** model to extract causal links

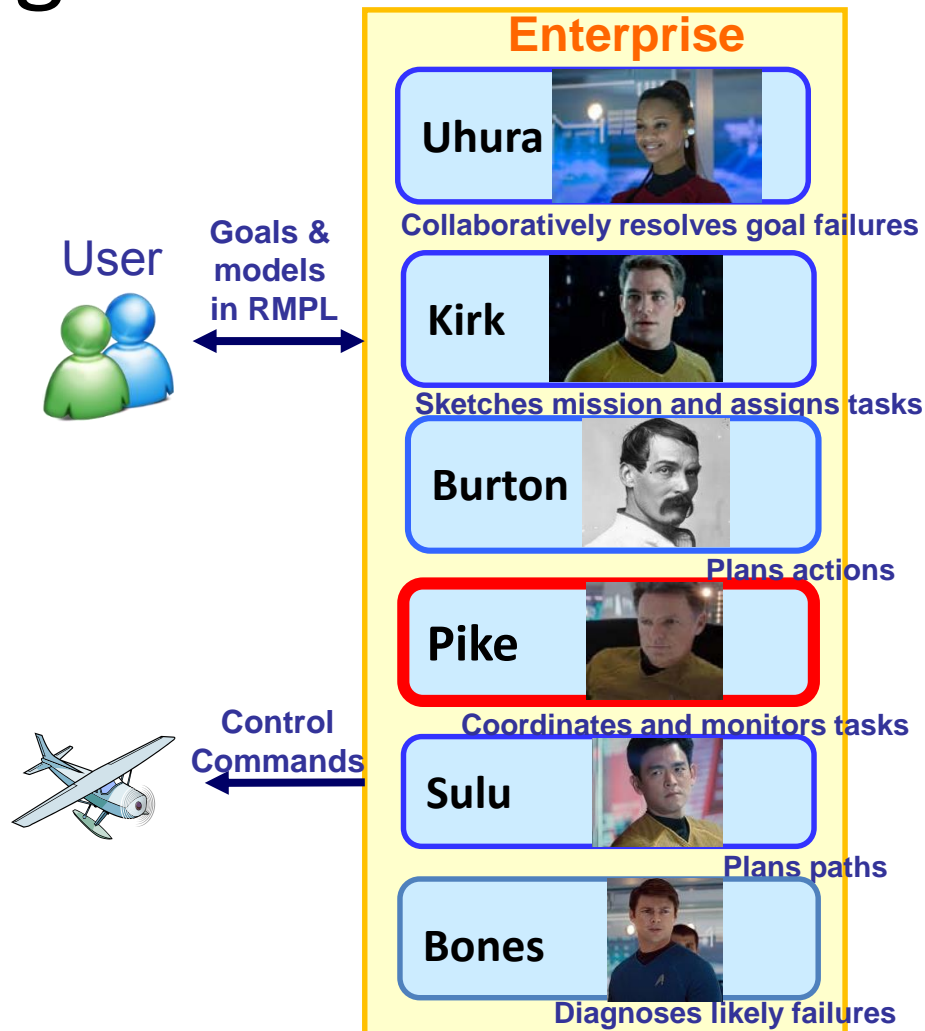
Online:

- Continuously **sense** monitor those causal links

Key takeaways

- Need to monitor the plan when executing
- Causal links:
 - What needs to be true, and when
 - Offline (checking) and online (monitoring)

A single “cognitive system” language and executive.



© source unknown. All rights reserved. This content is excluded from our Creative Commons license. For more information, see <https://ocw.mit.edu/help/faq-fair-use/>.

Causal link extraction algorithm

Extraction (non-temporal, totally-ordered plan - offline):

```

for each action in plan:
    for each precondition p of action:
        a_p = latest action in plan with effect p
        Add causal link: a_p to action over p
    
```

Monitoring (online):

```

while True:
    cls = currently active causal links (based on actions)
    state = measure state with sensors
    for each causal link cl in cls:
        p = predicate associated with cl
        if not(p in state):
            Trigger execution monitoring exception!
    
```

MIT OpenCourseWare
<https://ocw.mit.edu>

16.412J / 6.834J Cognitive Robotics
Spring 2016

For information about citing these materials or our Terms of Use, visit: <https://ocw.mit.edu/terms>.