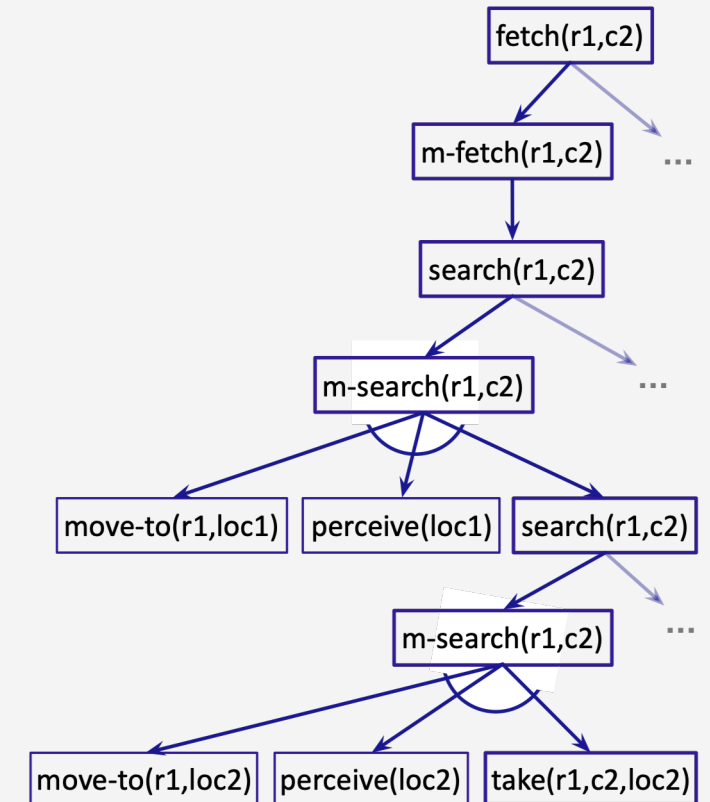


# Automated Planning and Acting Refinement Methods

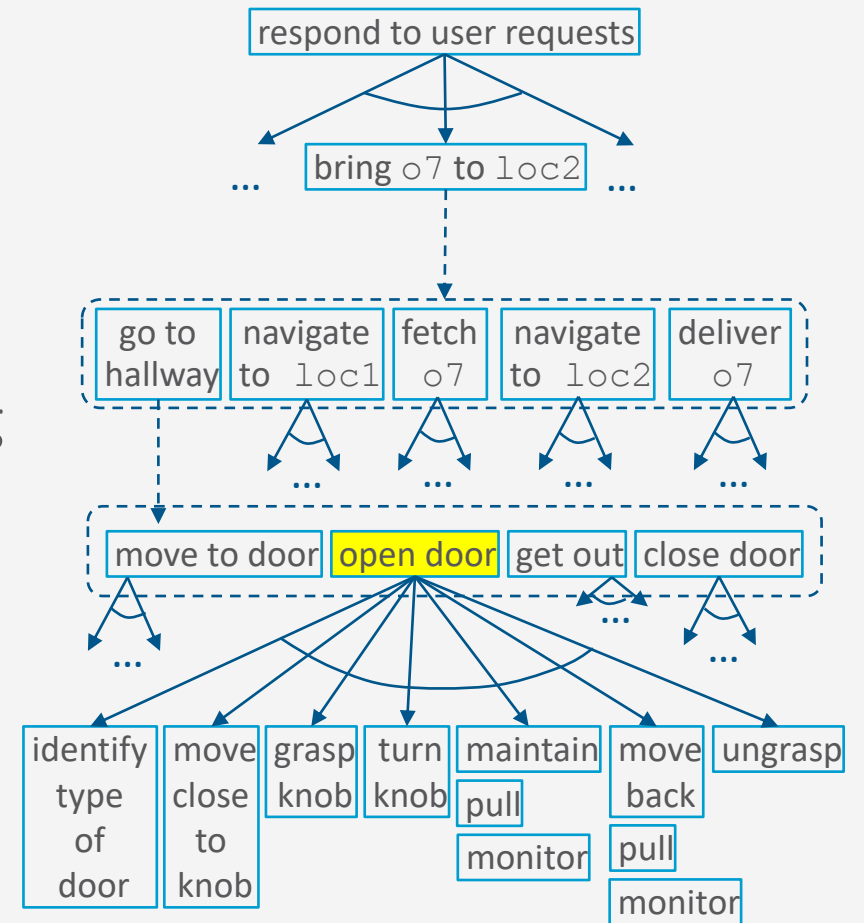


## Content: Planning and Acting

1. With **Deterministic Models**
2. With **Refinement Methods**
  - Operational Models
  - Refinement-Acting Machine
  - Refinement Planning
  - Acting and Refinement Planning
3. With **Temporal Models**
4. With **Nondeterministic Models**
5. With **Probabilistic Models**
6. By **Decision Making**
  - A. Foundations
  - B. Extensions
  - C. Structure
7. With **human-awareness**

# Motivation & Assumptions

- Hierarchically organised deliberation
  - At high levels, abstract actions
  - At lower levels, more detail
- Refine abstract actions into ways of carrying them out
  - How?
- Remove / weaken assumptions from classical planning
  - Characteristics
    - Dynamic environment
    - Imperfect information
    - Overlapping actions
    - Nondeterminism
    - Hierarchy
    - Discrete and continuous variables



Planning

Acting

# Outline per the Book

## 3.1 *Representation*

- State variables, commands, refinement methods
- Example

## 3.2 *Acting*

- RAE (Refinement Acting Engine)
- Example
- Extensions

## 3.3 *Planning*

- Motivation and basic ideas
- Deterministic action models
- SeRPE (Sequential Refinement Planning Engine)

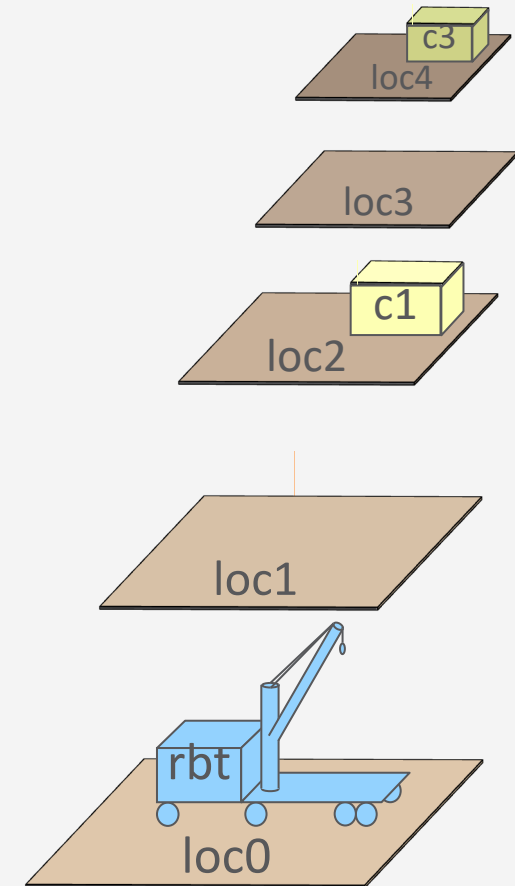
## 3.4 *Using Planning in Acting*

- Techniques
- Caveats



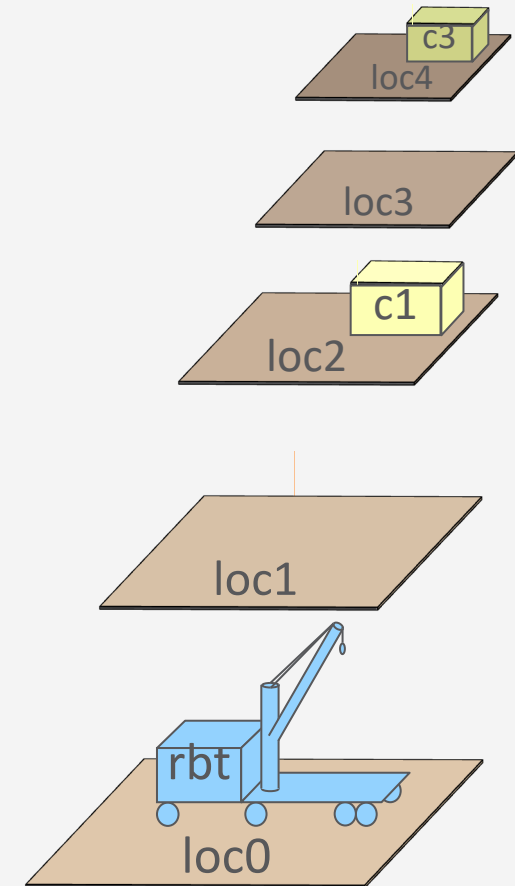
## State-variable Representation (Recap)

- Objects:
  - $Robots = \{rbt\}$
  - $Containers = \{c1, c2, c3, \dots\}$
  - $Locations = \{loc0, loc1, loc2, \dots\}$
- State variables: syntactic terms to which we can assign values
  - $loc(r) \in Locations$
  - $load(r) \in Containers \cup \{nil\}$
  - $pos(c) \in Locations \cup Robots \cup \{unknown\}$
  - $view(r, l) \in \{T, F\}$ 
    - whether robot  $r$  has looked at location  $l$
    - $r$  can only see what is at its current location
- State: assign a value to each state variable
  - $\{loc(rbt) = loc0, pos(c1) = loc2, pos(c3) = loc4, pos(c2) = unknown, \dots\}$



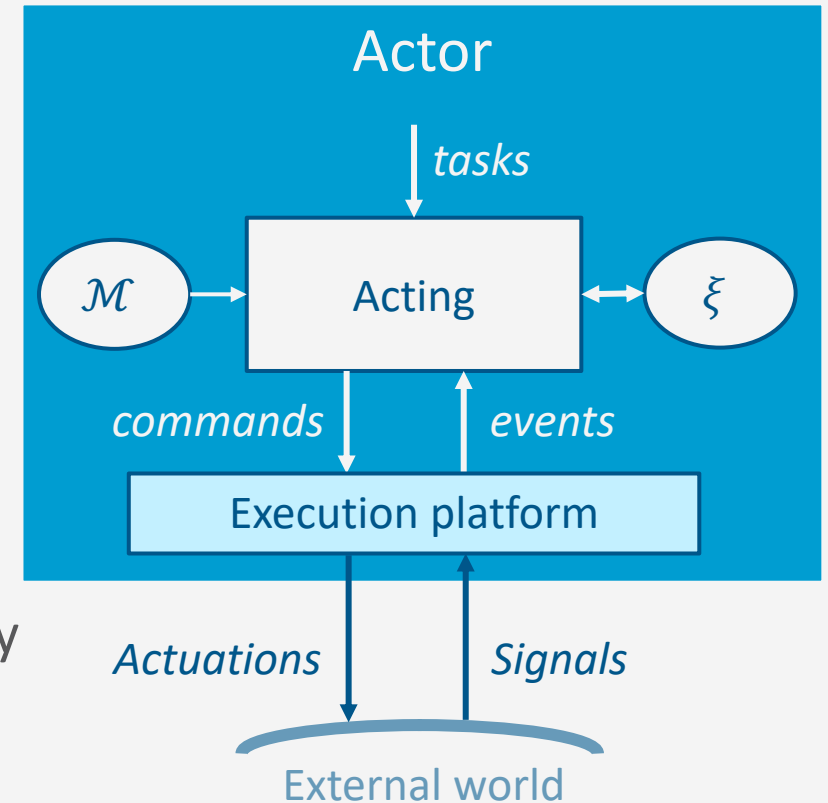
## State-variable Representation: Extensions

- Range  $\text{ran}(x)$ 
  - Can be finite, infinite, continuous, discontinuous, vectors, matrices, other data structures
- Assignment statement  $x \leftarrow \text{expr}$ 
  - Expression  $\text{expr}$  returns a ground value in  $\text{ran}(x)$  and has no side-effects on the current state
- Tests (e.g., preconditions)
  - Simple:  $x = v, x \neq v, x > v, x < v$
  - Compound: conjunction, disjunction, or negation of simple tests



## Commands

- **Command**: primitive function that the execution platform can perform
  - $take(r, o, l)$ : robot  $r$  takes object  $o$  at location  $l$
  - $put(r, o, l)$ :  $r$  puts  $o$  at location  $l$
  - $perceive(r, l)$ : robot  $r$  perceives what objects are at  $l$ 
    - $r$  can only perceive what is at its current location
- **Event**: occurrence detected by execution platform
  - $event-name(args)$
  - Exogenous changes in the environment to which the actor may have to react
  - E.g., emergency signal, arrival of transportation vehicle
- For later:  $\mathcal{M}$ : library of methods,  $\xi$ : current state (abstraction)



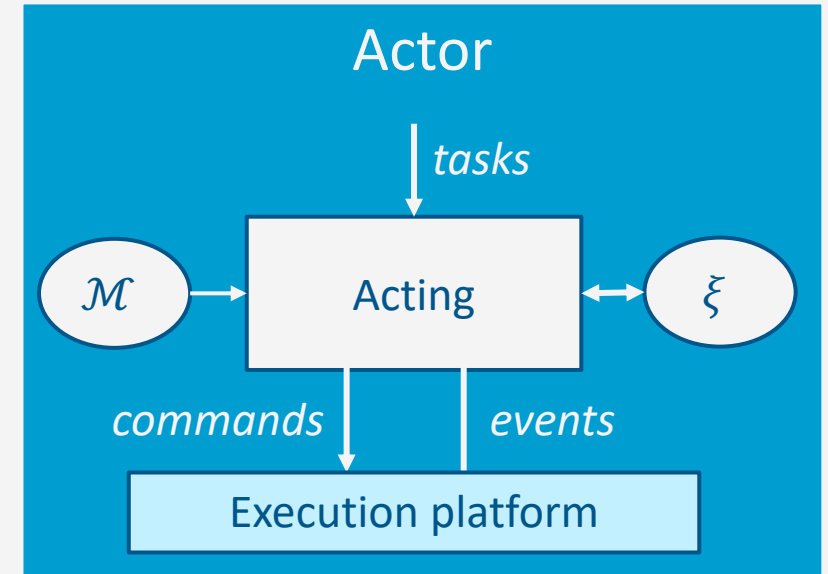
## Tasks and Methods

- **Task**: an activity for the actor to perform
  - Could be an abstract action of a plan
- For each task, a set of **refinement methods**
  - **Operational** models:
    - Tell *how* to perform the task
    - Do not predict *what* it will do

```

method-name(arg1, ..., argk)
task: task-identifier
pre: test
body: a program
  
```

- assignment statements
- control constructs: if-then-else, while, ...
- tasks (can extend to include events, goals)
- commands to the execution platform



## Example: “open door” task



- What kind:
  - Hinged on left
  - Opens toward us
  - Lever handle

→ Refinement method

```

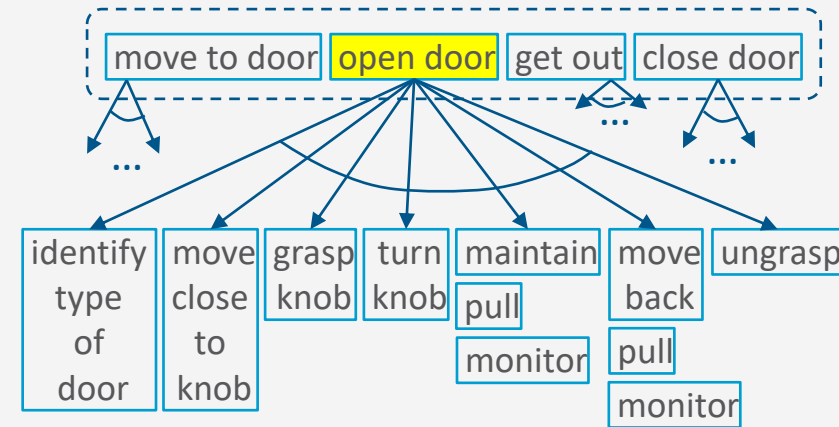
m-opendoor(r,d,l,h)
  task: opendoor(r,d)
  pre:  loc(r) = l ∧ adj(l,d)
        ∧ handle(d,h)
  body:
    while ¬reachable(r,h) do
      move-close(r,h)
      monitor-status(r,d)
      if door-status(d)=closed then
        unlatch(r,d)
        throw-wide(r,d)
      end-monitor-status(r,d)
  
```

```

m1-unlatch(r,d,l,o)
  task: unlatch(r,d)
  pre:  loc(r,l) ∧ toward-side(l,d) ∧
        side(d,left) ∧ type(d,rotate) ∧ handle(d,o)
  body: grasp(r,o)
        turn(r,o,alpha1)
        pull(r,val1)
        if door-status(d)=cracked then ungrasp(r,o)
        else fail
  
```

```

m1-throw-wide(r,d,l,o)
  task: throw-wide(r,d)
  pre:  loc(r,l) ∧ toward-side(l,d) ∧
        side(d,left) ∧ type(d,rotate) ∧
        handle(d,o) ∧ door-status(d)=cracked
  body: grasp(r,o)
        pull(r,val1)
        move-by(r,val2)
  
```



## Intermediate Summary

- 3.1 Operational models
  - Tasks, events
  - Commands to the execution platform
  - Extensions to state-variable representation
  - Refinement method: name, task/event, preconditions, body
  - Example: opening a door

# Outline per the Book

## 3.1 *Representation*

- State variables, commands, refinement methods
- Example

## 3.2 *Acting*

- RAE (Refinement Acting Engine)
- Example
- Extensions

## 3.3 *Planning*

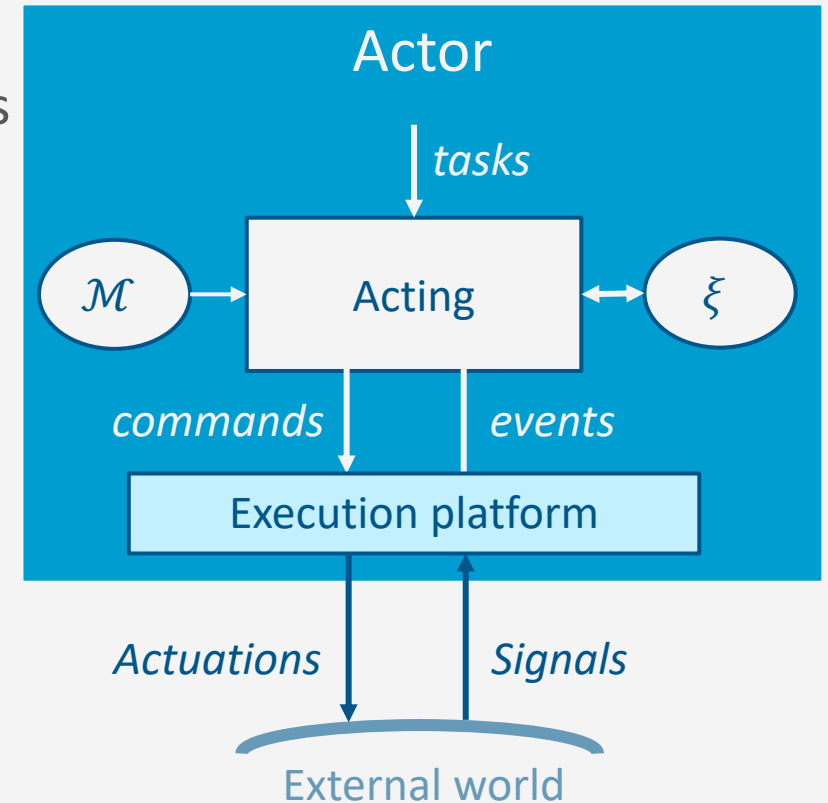
- Motivation and basic ideas
- Deterministic action models
- SeRPE (Sequential Refinement Planning Engine)

## 3.4 *Using Planning in Acting*

- Techniques
- Caveats

## RAE (Refinement Acting Engine)

- Based on OpenPRS programming language
  - Open-source robotics software, deployed in many applications
- Input
  - External tasks, events, current state  $\xi$ , library of methods  $\mathcal{M}$
- Output
  - Commands to execution platform
- Perform multiple tasks / events in parallel
  - Purely reactive, no lookahead
- For each task/event, a **refinement stack**
  - Current path in RAE's search tree for the task / event
- **Agenda** = {all current refinement stacks}





# RAE (Refinement Acting Engine)

- Input
  - Library of methods  $\mathcal{M}$
  - External tasks and events from an input stream, current state  $\xi$

## • Basic idea

### loop:

- **if** new external tasks/events **then**
  - Add them to Agenda
- **for each** stack in Agenda
  - Progress it
  - Remove it if it is finished

### RAE ( $\mathcal{M}$ )

*Agenda*  $\leftarrow \emptyset$

**loop**

**until** the input of external tasks and events is empty **do**

read  $\tau$  in the input stream

*Candidates*  $\leftarrow$  Instances( $\mathcal{M}, \tau, \xi$ )

**if** *Candidates* =  $\emptyset$  **then**

output("failed to address"  $\tau$ )

**else do**

arbitrarily choose  $m \in$  *Candidates*

*Agenda*  $\leftarrow$  *Agenda*  $\cup$   $\{(\tau, m, \text{nil}, \emptyset)\}$

**for** each *stack*  $\in$  *Agenda* **do**

Progress(*stack*)

**if** *stack* =  $\emptyset$  **then**

*Agenda*  $\leftarrow$  *Agenda*  $\setminus$  {*stack*}

### Stack element ( $\tau, m, i, \text{tried}$ )

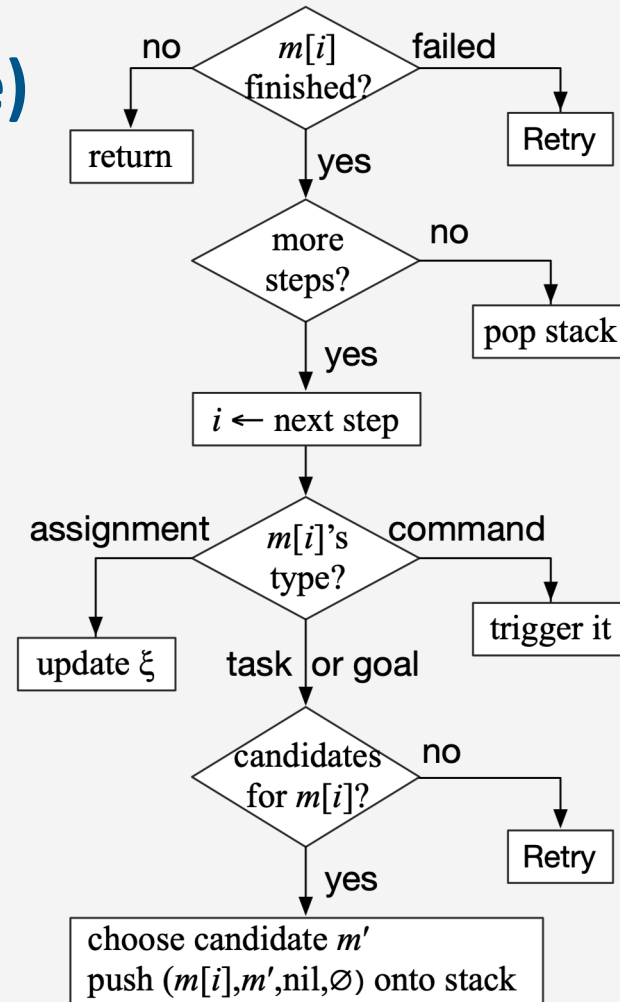
$\tau$  task

$m$  instance of a method in  $\mathcal{M}$

$i$  instruction pointer to  
step in body of  $m$

*tried* method instances already tried

# Progress (subroutine)



Just a decision tree

## Progress (stack)

```

    (τ, m, i, tried) ← top(stack)
    if i ≠ nil and m[i] is a command then
      case status(m[i])
        running: return
        failure: Retry(stack); return
        done: continue
    if i is the last step of m then
      pop(stack)
    else do
      i ← nextstep(m, i)
      case type(m[i])
        assignment: update ξ according
                     to m[i]; return
        command: trigger m[i]; return
        task or goal: continue
    τ' ← m[i]
    Candidates ← Instances(M, τ', ξ)
    if Candidates = ∅ then
      Retry(stack)
    else do
      arbitrarily choose m' ∈ Candidates
      push((τ', m', nil, ∅), stack)
  
```

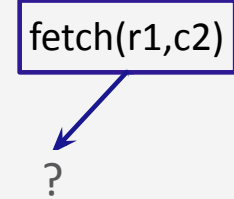
# Example

```
m-fetch(r,c)
  task: fetch(r,c)
  pre:
  body:
    if pos(c) = unknown then
      search(r,c)
    else if loc(r) = pos(c) then
      take(r,c,pos(c))
    else do
      move-to(r,pos(c))
      take(r,c,pos(c))
```

```
m-search(r,c)
  task: search(r,c)
  pre: pos(c) = unknown
  body:
    if  $\exists l$  (view(r,l) = F) then
      move-to(r,l)
      perceive(l)
      if pos(c) = l then
        take(r,c,l)
      else search(r,c)
    else fail
```

```
 $\tau$ : fetch(r1,c2)
m: ?
i: (see method)
tried: $\emptyset$ 
```

*Refinement stack*



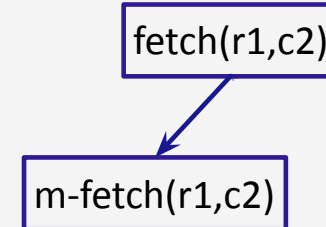
## Example

```
m-fetch(r,c)
  task: fetch(r,c)
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 $\tau$ : fetch(r1,c2)
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*Refinement stack*



```

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  task: fetch(r,c)
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    else do
      move-to(r,pos(c))
      take(r,c,pos(c))

```

```

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  task: search(r,c)
  pre: pos(c) = unknown
  body:
    if  $\exists l$  (view(r,l) = F) then
      move-to(r,l)
      perceive(l)
      if pos(c) = l then
        take(r,c,l)
      else search(r,c)
    else fail

```

```

 $\tau$ : search(r1,c2)
m: ?
i: (see method)
tried: $\emptyset$ 

```

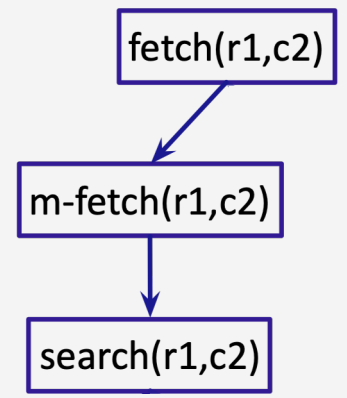
```

 $\tau$ : fetch(r1,c2)
m: m-fetch(r1,c2)
i: (see method)
tried: $\emptyset$ 

```

Refinement stack

# Example



## Example

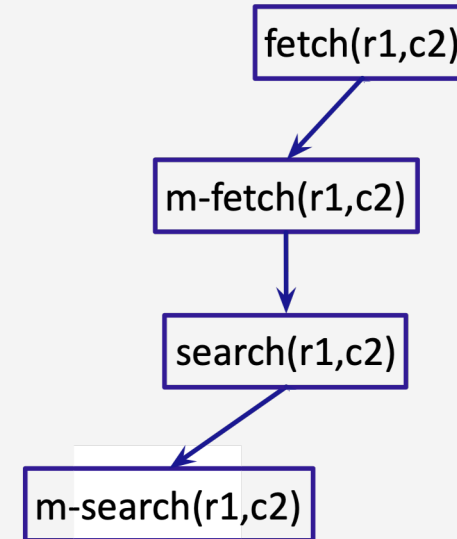
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      else search(r,c)
    else fail
```

```
 $\tau$ : search(r1,c2)
m: m-search(r1,c2)
i: (see method)
tried: $\emptyset$ 
```

```
 $\tau$ : fetch(r1,c2)
m: m-fetch(r1,c2)
i: (see method)
tried: $\emptyset$ 
```

Refinement stack



# Example

```
m-fetch(r,c)
  task: fetch(r,c)
  pre:
  body:
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    else do
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      take(r,c,pos(c))
```

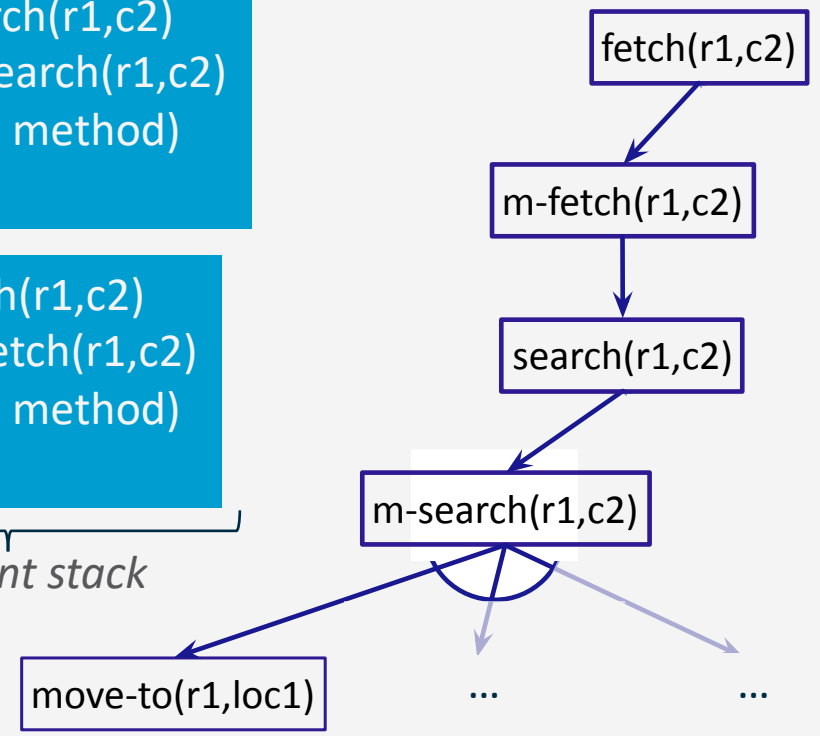
```
m-search(r,c)
  task: search(r,c)
  pre: pos(c) = unknown
  body:
    if  $\exists l$  (view(r,l) = F) then
      move-to(r,l)
      perceive(l)
      if pos(c) = l then
        take(r,c,l)
      else search(r,c)
    else fail
```

...

$\tau$ : search(r1,c2)  
 $m$ : m-search(r1,c2)  
 $i$ : (see method)  
 $tried:\emptyset$

$\tau$ : fetch(r1,c2)  
 $m$ : m-fetch(r1,c2)  
 $i$ : (see method)  
 $tried:\emptyset$

Refinement stack



# Example

```
m-fetch(r,c)
  task: fetch(r,c)
  pre:
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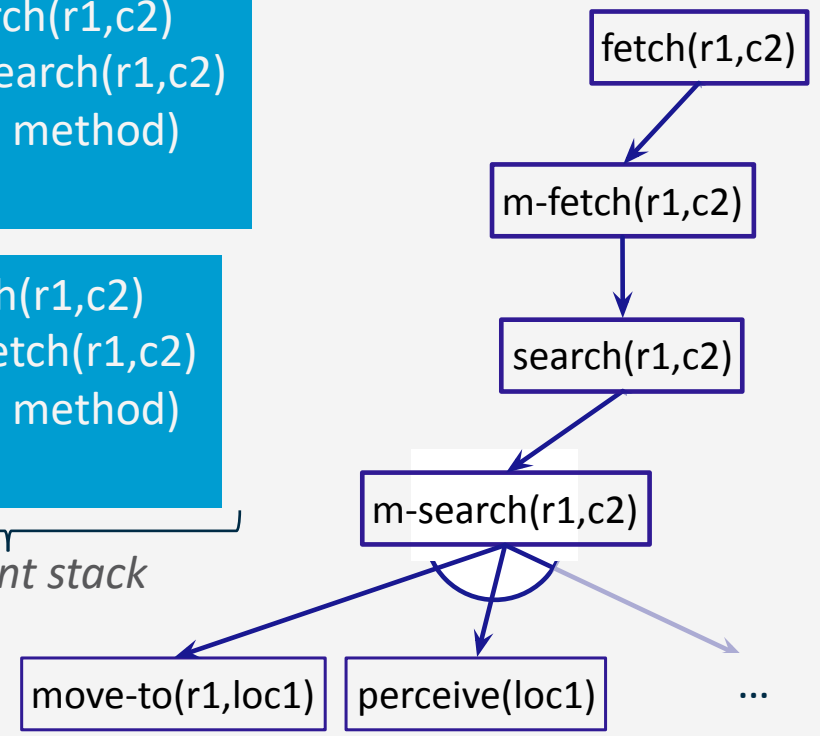
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  task: search(r,c)
  pre: pos(c) = unknown
  body:
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      move-to(r,l)
      perceive(l)
      if pos(c) = l then
        take(r,c,l)
      else search(r,c)
    else fail
```

...

```
 $\tau$ : search(r1,c2)
m: m-search(r1,c2)
i: (see method)
tried: $\emptyset$ 
```

```
 $\tau$ : fetch(r1,c2)
m: m-fetch(r1,c2)
i: (see method)
tried: $\emptyset$ 
```

Refinement stack





# Example

```
m-fetch(r,c)
  task: fetch(r,c)
  pre:
  body:
    if pos(c) = unknown then
      search(r,c)
    else if loc(r) = pos(c) then
      take(r,c,pos(c))
    else do
      move-to(r,pos(c))
      take(r,c,pos(c))
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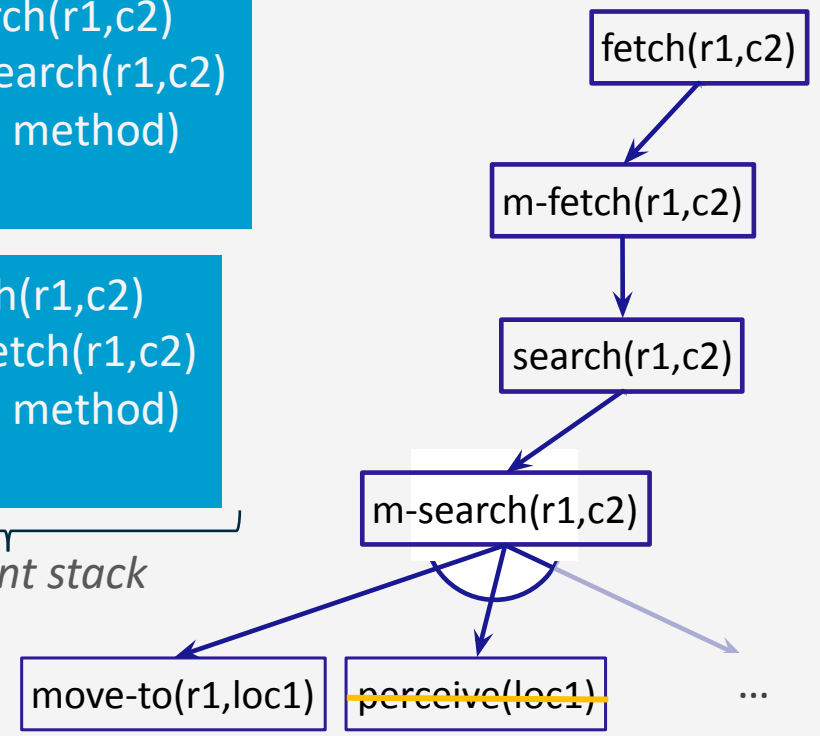
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  task: search(r,c)
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      perceive(l)
      if pos(c) = l then
        take(r,c,l)
      else search(r,c)
    else fail
```

...

```
 $\tau$ : search(r1,c2)
m: m-search(r1,c2)
i: (see method)
tried: $\emptyset$ 
```

```
 $\tau$ : fetch(r1,c2)
m: m-fetch(r1,c2)
i: (see method)
tried: $\emptyset$ 
```

Refinement stack



sensor failure

# Example

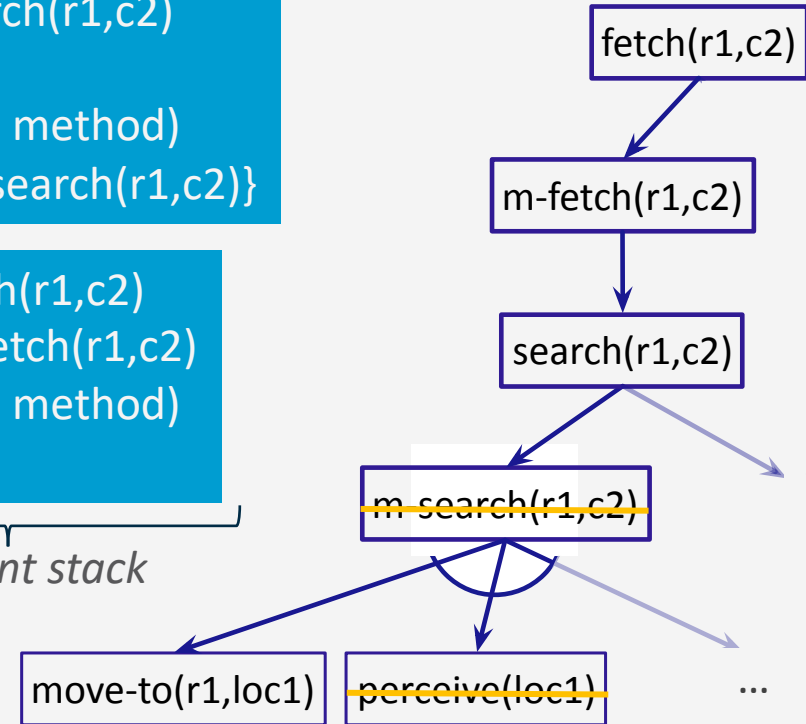
```
m-fetch(r,c)
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```
m-search(r,c)
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  body:
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      move-to(r,l)
      perceive(l)
      if pos(c) = l then
        take(r,c,l)
      else search(r,c)
    else fail
```

```
 $\tau$ : search(r1,c2)
m: ?
i: (see method)
tried: {m-search(r1,c2)}
```

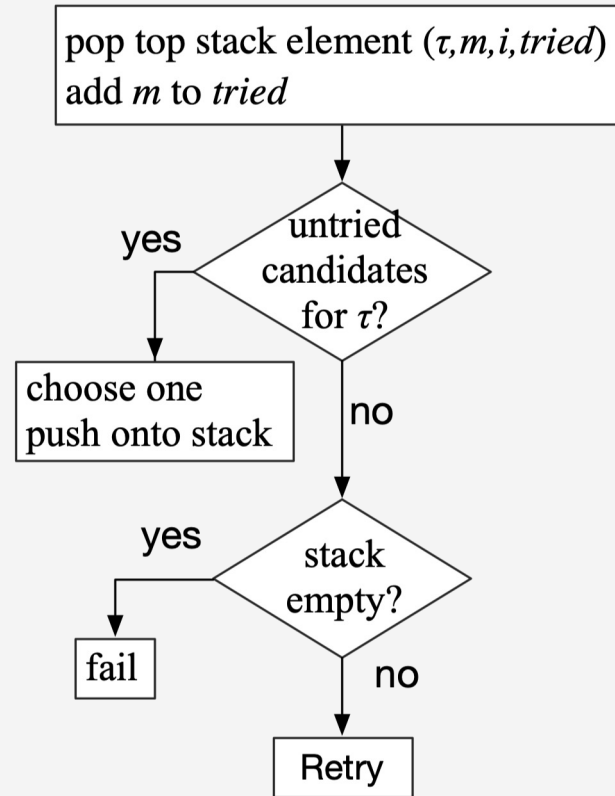
```
 $\tau$ : fetch(r1,c2)
m: m-fetch(r1,c2)
i: (see method)
tried:  $\emptyset$ 
```

Refinement stack



If other candidates for *search(r1,c2)*, try them.

# Retry (subroutine)



Another decision tree

## Retry(*stack*)

```

(τ, m, i, tried) ← pop(stack)
tried ← tried ∪ {m}
Candidates ← Instances(M, τ, ξ) \ tried
if Candidates ≠ ∅ then
  arbitrarily choose m' ∈ Candidates
  push((τ, m', nil, ∅), stack)
else do
  if stack ≠ ∅ then
    Retry(stack)
  else do
    output("failed to accomplish" τ)
    Agenda ← Agenda \ stack
  
```

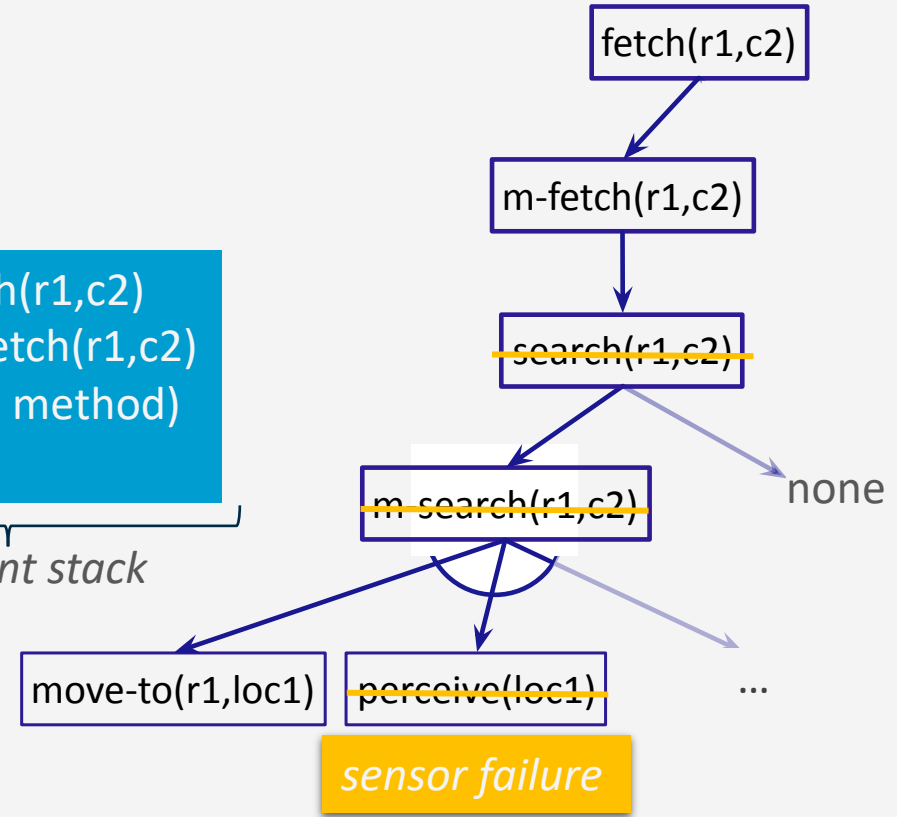
# Example

```
m-fetch(r,c)
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  pre:
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    else do
      move-to(r,pos(c))
      take(r,c,pos(c))
```

```
m-search(r,c)
  task: search(r,c)
  pre: pos(c) = unknown
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    if  $\exists l$  (view(r,l) = F) then
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      else search(r,c)
    else fail
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```
 $\tau$ : fetch(r1,c2)
 $m$ : m-fetch(r1,c2)
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 $tried:\emptyset$ 
```

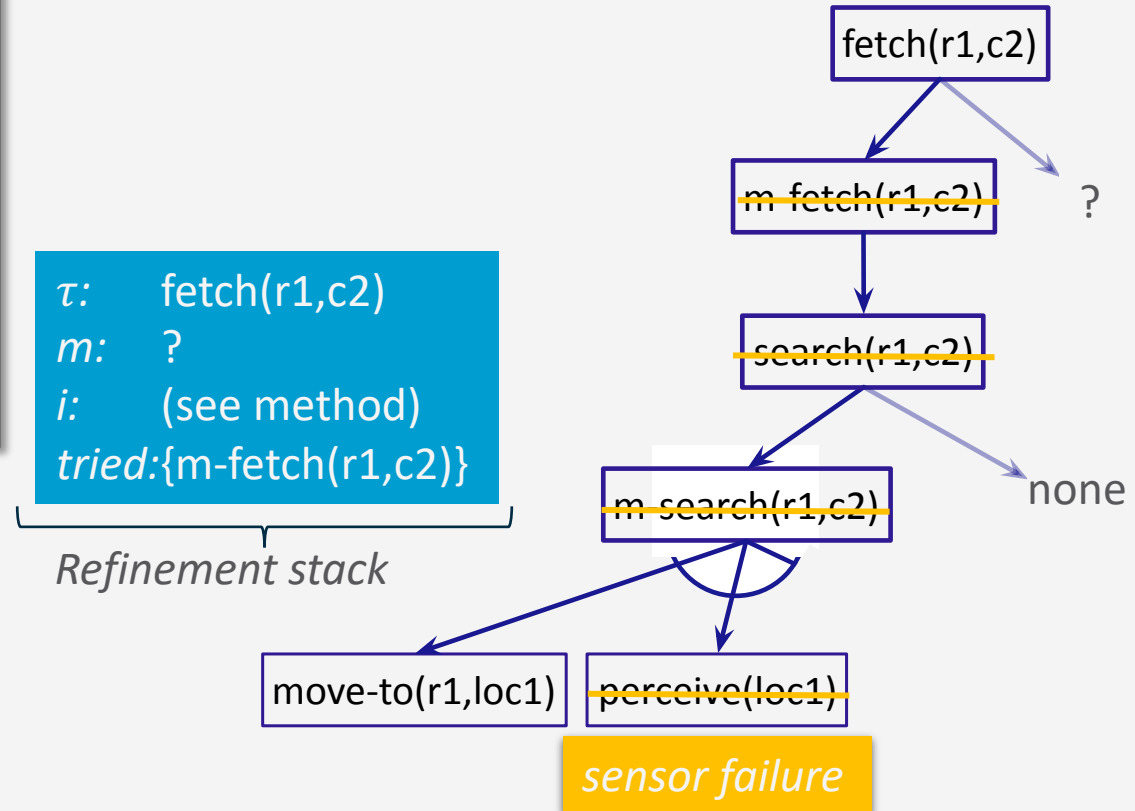
Refinement stack



# Example

```
m-fetch(r,c)
  task: fetch(r,c)
  pre:
  body:
    if pos(c) = unknown then
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    else if loc(r) = pos(c) then
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    else do
      move-to(r,pos(c))
      take(r,c,pos(c))
```

```
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  task: search(r,c)
  pre: pos(c) = unknown
  body:
    if  $\exists l$  (view(r,l) = F) then
      move-to(r,l)
      perceive(l)
      if pos(c) = l then
        take(r,c,l)
      else search(r,c)
    else fail
```



If other candidates for *fetch(r1,c2)*, try them.

## Extensions to RAE: Events

- Example: an emergency
  - If  $r$  is not already handling another emergency, then
    - Stop what it is doing
    - Go handle the emergency

m-emergency( $r, l, i$ )

event: emergency( $l, i$ )

pre: emergency-handling( $r$ ) = F

body: emergency-handling( $r$ )  $\leftarrow$  T

if load( $r$ )  $\neq$  nil then

    put( $r$ , load( $r$ ))

    move-to( $l$ )

    address-emergency( $l, i$ )

RAE ( $\mathcal{M}$ )

Agenda  $\leftarrow \emptyset$

loop

**until** the input of external tasks and  
     events is empty **do**

    read  $\tau$  in the input stream

    Candidates  $\leftarrow$  Instances( $\mathcal{M}, \tau, \xi$ )

**if** Candidates =  $\emptyset$  **then**

        output("failed to address"  $\tau$ )

**else do**

        arbitrarily choose  $m \in$  Candidates

        Agenda  $\leftarrow$  Agenda  $\cup \{(\tau, m, nil, \emptyset)\}$

**for** each stack  $\in$  Agenda **do**

        Progress(stack)

**if** stack =  $\emptyset$  **then**

            Agenda  $\leftarrow$  Agenda  $\setminus$  {stack}

method-name( $arg_1, \dots, arg_k$ )

task: event-identifier

pre: test

body: a program

## Extensions to RAE: Goals

- Write as a special task
  - Like others, but includes monitoring: modify **Progress**
    - if *condition* becomes true before finishing *body(m)*, stop early
    - if *condition* isn't true after finishing *body(m)*, fail and try another method

```
m-fetch(r,c)
  task: fetch(r,c)
  pre:
  body: if pos(c) = unknown then
        search(r,c)
        achieve(pos(c)≠unknown)
        move-to(r,pos(c))
        take(r,c,pos(c))
```

```
m-find-where(r,c)
  task: achieve(pos(c)≠unknown)
  pre:
  body: while exists loc. l s.t. view(l) = F do
        move-to(l)
        perceive(l)
```

```
RAE ( $\mathcal{M}$ )
  Agenda  $\leftarrow \emptyset$ 
  loop
    until the input of external tasks and
           events is empty do
      read  $\tau$  in the input stream
      Candidates  $\leftarrow$  Instances( $\mathcal{M}, \tau, \xi$ )
      if Candidates =  $\emptyset$  then
        output("failed to address"  $\tau$ )
      else do
        arbitrarily choose  $m \in$  Candidates
        Agenda  $\leftarrow$  Agenda  $\cup \{(\tau, m, nil, \emptyset)\}$ 
      for each stack  $\in$  Agenda do
        Progress(stack)
        if stack =  $\emptyset$  then
          Agenda  $\leftarrow$  Agenda  $\setminus \{stack\}$ 
```

```
method-name(arg1, ..., argk)
  task: achieve(condition)
  pre: test
  body: a program
```

## Other Extensions to RAE

- Concurrent subtasks:
  - Refinement stack for each one
- Controlling the progress of tasks:
  - E.g., suspend a task for a while
  - If there are multiple stacks, which ones get higher priority?
    - Application-specific heuristics
- For a task  $\tau$ , which candidate to try first?
  - **Refinement planning**

body of a method

...

{concurrent:  $\tau_1, \tau_2, \dots, \tau_n$ }

...

*Agenda* = {*stack*<sub>1</sub>, *stack*<sub>2</sub>, ..., *stack*<sub>n</sub>}

*Candidates*  $\leftarrow$  Instances( $\mathcal{M}, \tau, \xi$ )



## Intermediate Summary

- 3.2 Refinement Acting Engine (RAE)
  - Purely reactive: select a method and apply it
  - RAE: input stream, *Candidates*, *Instances*, *Agenda*, refinement stacks
  - Progress: command status, nextstep, type of step
  - Retry: *Candidates \ tried*, *Agenda \ stack*
  - Refinement trees
  - Concurrent tasks: for each, a refinement stack
  - Goal: achieve(condition), uses monitoring
  - Controlling progress, heuristics

# Outline per the Book

## 3.1 *Representation*

- State variables, commands, refinement methods
- Example

## 3.2 *Acting*

- RAE (Refinement Acting Engine)
- Example
- Extensions

## 3.3 *Planning*

- Motivation and basic ideas
- Deterministic action models
- SeRPE (Sequential Refinement Planning Engine)

## 3.4 *Using Planning in Acting*

- Techniques
- Caveats

## Motivation

- When dealing with an event or task, RAE may need to make either/or choices
  - *Agenda*: tasks  $\tau_1, \tau_2, \dots, \tau_n$ 
    - Several tasks/events, how to prioritise?
  - Candidates for  $\tau_1$ :  $m_1, m_2, \dots$ 
    - Several candidate methods or commands, which one to try first?
- RAE immediately executes commands
  - Bad choices may be **costly** or **irreversible**

# Refinement Planning

- Basic idea:
  - Go step by step through RAE, but do not send commands to execution platform
  - For each command, use a descriptive action model to predict the next state
    - Tells *what*, not *how*
  - Whenever we need to choose a method
    - Try various possible choices, explore consequences, choose best
- Generalisation of Hierarchical Task Network (HTN) planning
  - HTN planning: body of a method is a list of tasks
  - Here: body of method is the same program RAE uses
  - Use it to *generate* a list of tasks

# SeRPE (Sequential Refinement Planning Engine)

- SeRPE inputs

```

 $\mathcal{M} = \{\text{methods}\}$ 
 $\mathcal{A} = \{\text{action models}\}$ 
 $s = \text{initial state}$ 
 $\tau = \text{task or goal}$ 

```

- Which candidate method for  $\tau$ ?

- SeRPE: Nondeterministic choice

- Backtracking point
- How to implement?
  - Hierarchical adaptation of backtracking, A\*, GBFS, ...

- RAE: Arbitrary choice

- No search, purely reactive

```

SeRPE ( $\mathcal{M}, \mathcal{A}, s, \tau$ )
  Candidates  $\leftarrow$  Instances ( $\mathcal{M}, \tau, s$ )
  if Candidates =  $\emptyset$  then
    return failure
  nondeterministically choose  $m \in$  Candidates
  return Progress-to-finish ( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )

```

```

RAE ( $\mathcal{M}$ )
  Agenda  $\leftarrow \emptyset$ 
  loop
    until the input of external tasks and
           events is empty do
      read  $\tau$  in the input stream
      Candidates  $\leftarrow$  Instances ( $\mathcal{M}, \tau, \xi$ )
      if Candidates =  $\emptyset$  then
        output("failed to address"  $\tau$ )
      else do
        arbitrarily choose  $m \in$  Candidates
        Agenda  $\leftarrow$  Agenda  $\cup \{(\tau, m, \text{nil}, \emptyset)\}$ 
      for each stack  $\in$  Agenda do
        Progress(stack)
        if stack =  $\emptyset$  then
          Agenda  $\leftarrow$  Agenda  $\setminus \{\text{stack}\}$ 

```

# SeRPE (Sequential Refinement Planning Engine)

- SeRPE
  - One external task
  - Simulate progressing it all the way to the end
- RAE
  - Several external tasks
  - Each time through loop, progress each one by one step

```

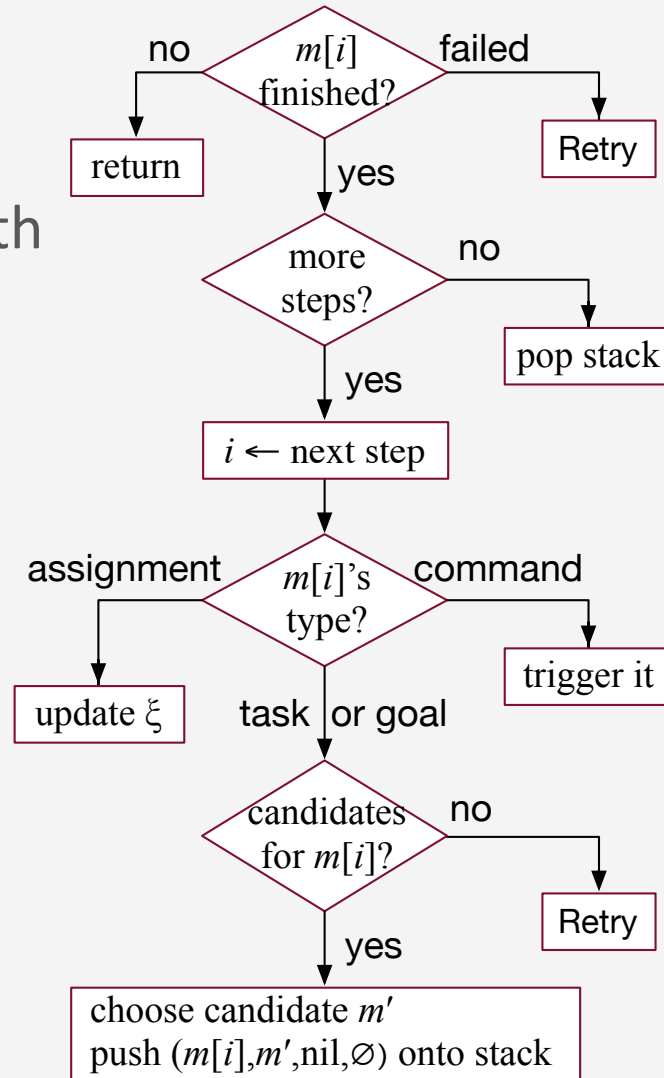
SeRPE ( $\mathcal{M}, \mathcal{A}, s, \tau$ )
  Candidates  $\leftarrow$  Instances ( $\mathcal{M}, \tau, s$ )
  if Candidates =  $\emptyset$  then
    return failure
  nondeterministically choose  $m \in$  Candidates
  return Progress-to-finish ( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )
  
```

```

RAE ( $\mathcal{M}$ )
  Agenda  $\leftarrow$   $\emptyset$ 
  loop
    until the input of external tasks and
      events is empty do
      read  $\tau$  in the input stream
      Candidates  $\leftarrow$  Instances ( $\mathcal{M}, \tau, \xi$ )
      if Candidates =  $\emptyset$  then
        output("failed to address"  $\tau$ )
      else do
        arbitrarily choose  $m \in$  Candidates
        Agenda  $\leftarrow$  Agenda  $\cup$   $\{(\tau, m, \text{nil}, \emptyset)\}$ 
      for each stack  $\in$  Agenda do
        Progress (stack)
        if stack =  $\emptyset$  then
          Agenda  $\leftarrow$  Agenda  $\setminus$  {stack}
  
```

## Progress-to-finish

- Like RAE progress with a loop around it
- Simulates the commands



Progress-to-finish ( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )

$i \leftarrow \text{nil}; \pi \leftarrow \langle \rangle$

loop

if  $\tau$  is a goal and  $s \models \tau$  then  
 return  $\pi$

if  $i$  is the last step of  $m$  then  
 if  $\tau$  is a goal and  $s \not\models \tau$  then  
 return failure

return  $\pi$

$i \leftarrow \text{nextstep}(m, i)$

case type( $m[i]$ )

assignment:

update  $s$  according to  $m[i]$

command:

$a \leftarrow \text{descriptive model of } m[i] \text{ in } \mathcal{A}$

if  $s \models \text{pre}(a)$  then

$s \leftarrow \gamma(s, a); \pi \leftarrow \pi.a$

else

return failure

task or goal:

$\pi' \leftarrow \text{SeRPE}(\mathcal{M}, \mathcal{A}, s, m[i])$

if  $\pi' = \text{failure}$  then

return failure

$s \leftarrow \gamma(s, \pi'); \pi \leftarrow \pi. \pi'$

## Progress-to-finish

- Inputs:  $\mathcal{M} = \{\text{methods}\}$ ,  $\mathcal{A} = \{\text{action models}\}$ ,  
 $s = \text{initial state}$ ,  $\tau = \text{task or goal}$ ,  
 $m = \text{chosen method}$
- Simulate RAE's goal monitoring
- If  $m[i]$  is a command:  
Use action model to predict outcome
- If current step is a task: Call SeRPE recursively
  - Recursion stack  $\approx$  RAE's refinement stack
- For failures, no Retry (RAE)
  - A failure means SeRPE could not find a solution
  - Implementation: hierarchical adaptations of backtracking, A\*, GBFS, ...

```

Progress-to-finish ( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )
   $i \leftarrow \text{nil}; \pi \leftarrow \langle \rangle$ 
  loop
    if  $\tau$  is a goal and  $s \models \tau$  then
      return  $\pi$ 
    if  $i$  is the last step of  $m$  then
      if  $\tau$  is a goal and  $s \not\models \tau$  then
        return failure
      return  $\pi$ 
     $i \leftarrow \text{nextstep}(m, i)$ 
    case type( $m[i]$ )
      assignment:
        update  $s$  according to  $m[i]$ 
      command:
         $a \leftarrow \text{descriptive model of } m[i] \text{ in } \mathcal{A}$ 
        if  $s \models \text{pre}(a)$  then
           $s \leftarrow \gamma(s, a); \pi \leftarrow \pi.a$ 
        else
          return failure
      task or goal:
         $\pi' \leftarrow \text{SeRPE}(\mathcal{M}, \mathcal{A}, s, m[i])$ 
        if  $\pi' = \text{failure}$  then
          return failure
         $s \leftarrow \gamma(s, \pi'); \pi \leftarrow \pi. \pi'$ 

```



*task*  
 put-in-pile( $c_1, p_2$ )

## Example

~~Candidates = {m1-put-in-pile( $c_1, p_2$ ),  
 m2-put-in-pile( $r, c_1, p_1, d, p', d'$ )}~~

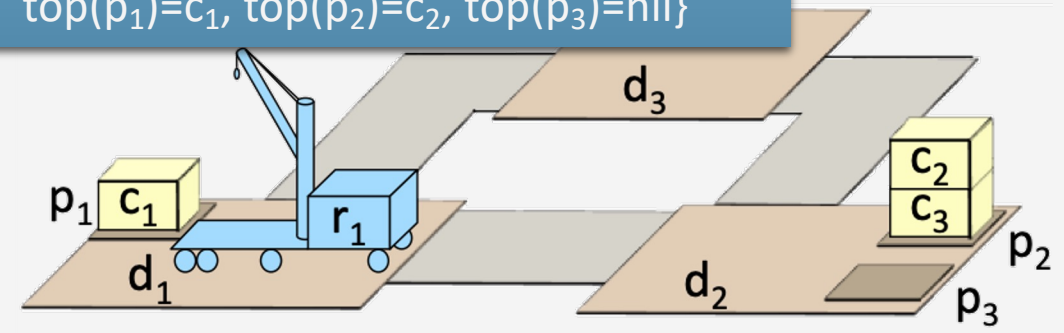
```

SeRPE ( $\mathcal{M}, \mathcal{A}, s, \tau$ )
  Candidates  $\leftarrow$  Instances ( $\mathcal{M}, \tau, s$ )
  if Candidates =  $\emptyset$  then
    return failure
  nondeterministically choose  $m \in$  Candidates
  return Progress-to-finish ( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )
  
```

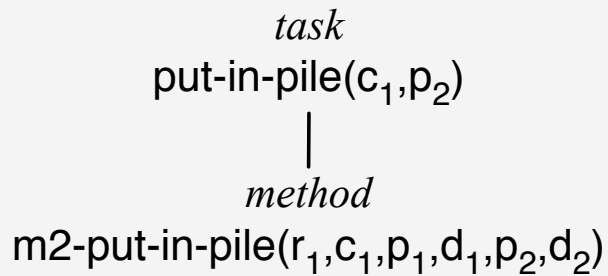
m1-put-in-pile( $c, p'$ )  
 task: put-in-pile( $c, p'$ )  
 pre: pile( $c$ )= $p'$   
 body: // empty

m2-put-in-pile( $r, c, p, d, p', d'$ )  
 task: put-in-pile( $c, p'$ )  
 pre: pile( $c$ )= $p \wedge$  at( $p, d$ )  $\wedge$  at( $p', d'$ )  
 $\wedge p \neq p' \wedge$  cargo( $r$ )=nil  
 body: **if** loc( $r$ )  $\neq$   $d$  **then** navigate( $r, d$ )  
 uncover( $c$ )  
 load( $r, c, \text{pos}(c), p, d$ )  
**if** loc( $r$ )  $\neq$   $d'$  **then** navigate( $r, d'$ )  
 unload( $r, c, \text{top}(p'), p', d$ )

$s_0 = \{\text{loc}(r_1)=d_1, \text{cargo}(r_1)=\text{nil}, \text{occupied}(d_1)=T,$   
 $\text{occupied}(d_2)=F, \text{occupied}(d_3)=F,$   
 $\text{pos}(c_1)=\text{nil}, \text{pos}(c_2)=c_3, \text{pos}(c_3)=\text{nil},$   
 $\text{pile}(c_1)=p_1, \text{pile}(c_2)=p_2, \text{pile}(c_3)=p_2,$   
 $\text{top}(p_1)=c_1, \text{top}(p_2)=c_2, \text{top}(p_3)=\text{nil}\}$



# Example



## Refinement tree

- The SeRPE pseudocode does not return this, but can easily be modified to do so

## SeRPE ( $\mathcal{M}, \mathcal{A}, s, \tau$ )

*Candidates*  $\leftarrow$  Instances( $\mathcal{M}, \tau, s$ )

**if** *Candidates* =  $\emptyset$  **then**

**return** failure

nondeterministically choose  $m \in$  *Candidates*

**return** Progress-to-finish( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )

$r_1, c_1, p_1, d_1, p_2, d_2$

m2-put-in-pile( $r, c, p, d, p', d'$ )

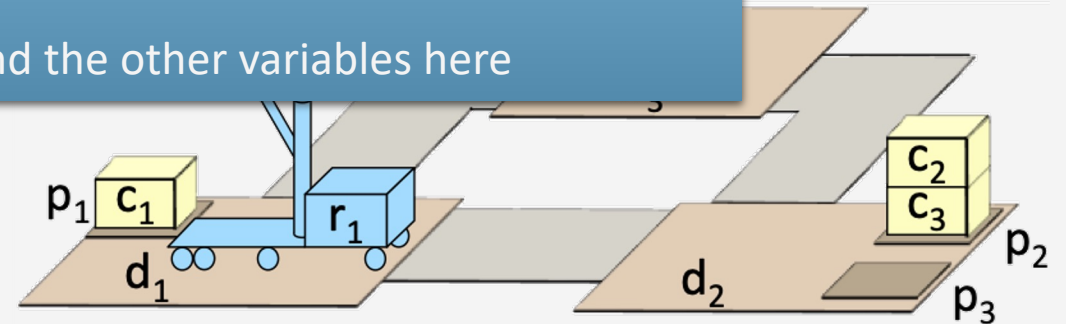
task: put-in-pile( $c, p'$ )

pre: pile( $c$ )= $p \wedge$  at( $p, d$ )  $\wedge$  at( $p', d'$ )  
 $\wedge p \neq p' \wedge$  cargo( $r$ )=nil

body: if loc( $r$ )  $\neq d$  then navigate( $r, d$ )  
uncover( $c$ )  
load( $r, c, \text{pos}(c), p, d$ )  
if loc( $r$ )  $\neq d'$  then navigate( $r, d'$ )  
unload( $r, c, \text{top}(p'), p', d$ )

- m2-put-in-pile starts with  $c=c_1, p'=p_2$ , and  $r, d, p', d'$  unbound

- Bind the other variables here



# Example

*task*  
put-in-pile( $c_1, p_2$ )  
|  
*method*  
m2-put-in-pile( $r_1, c_1, p_1, d_1, p_2, d_2$ )

## Progress-to-finish ( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )

```

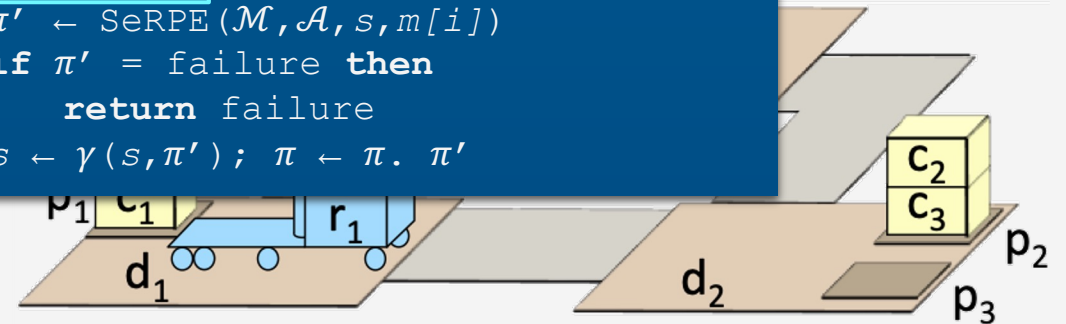
i ← nil; π ← ⟨⟩
loop
  if τ is a goal and s ⊨ τ then
    return π
  if i is the last step of m then
    if τ is a goal and s ⊭ τ then
      return failure
    return π
  i ← nextstep(m, i)
  case type(m[i])
  assignment:
    update s according to m[i]
  command:
    a ← descriptive model of m[i] in A
    if s ⊨ pre(a) then
      s ← γ(s, a); π ← π.a
    else
      return failure
  task or goal:
    π' ← SeRPE(ℳ, A, s, m[i])
    if π' = failure then
      return failure
    s ← γ(s, π'); π ← π. π'

```

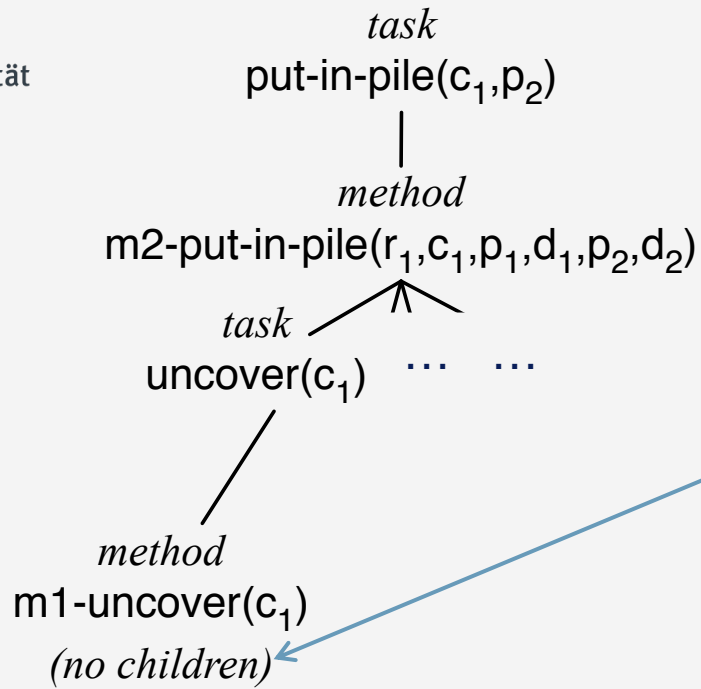
Refinement

loc( $r_1$ ) =  $d_1 = d$

$r_1, c_1, p_1, d_1, p_2, d_2$   
m2-put-in-pile( $r, c, p, d, p', d'$ )  
task: put-in-pile( $c, p'$ )  
pre: pile( $c$ )= $p \wedge$  at( $p, d$ )  $\wedge$  at( $p', d'$ )  
 $\wedge p \neq p' \wedge$  cargo( $r$ )= $nil$   
body: if loc( $r$ )  $\neq d$  then navigate( $r, d$ )  
uncover( $c$ )  
load( $r, c, pos(c), p, d$ )  
if loc( $r$ )  $\neq d'$  then navigate( $r, d'$ )  
unload( $r, c, top(p'), p', d$ )



# Example



```

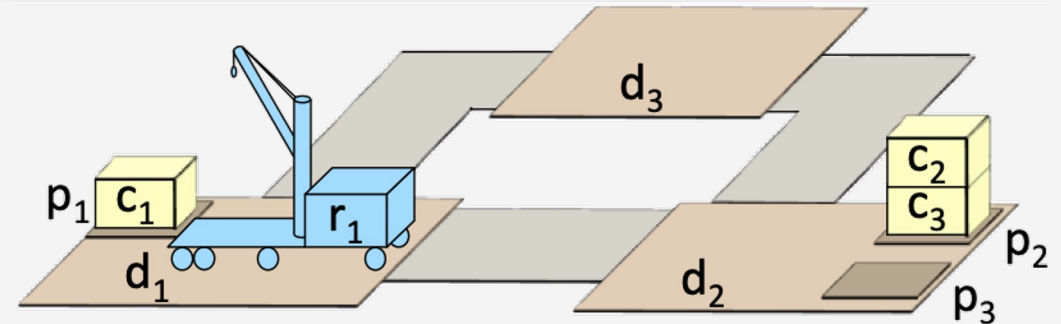
m1-uncover(c)
  task: uncover(c)
  pre: top(pile(c))=c
  body: // empty
  
```

```

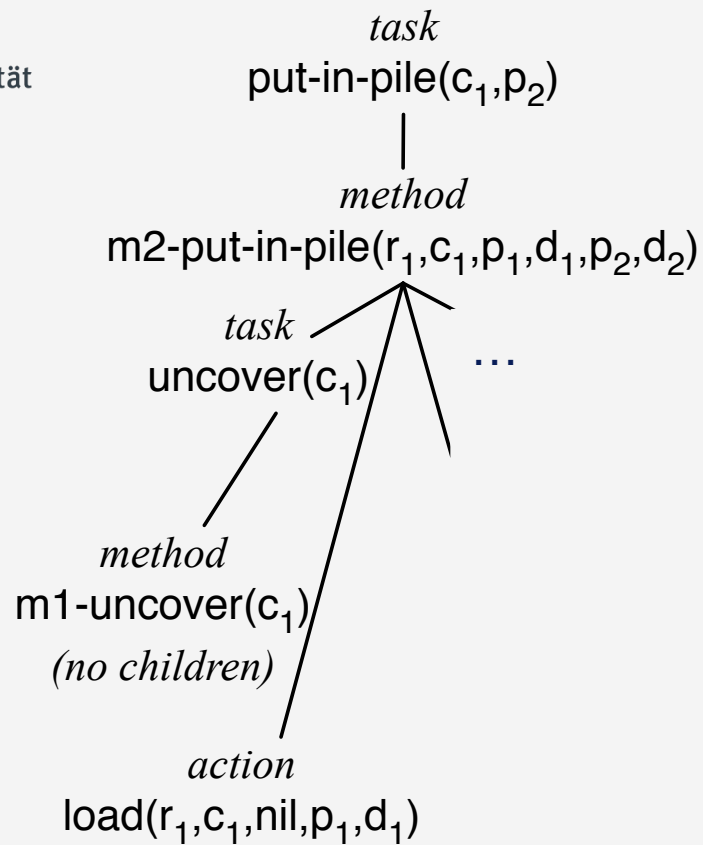
m2-uncover(r, c, c', p', d)
  task: uncover(c)
  pre: pile(c)=p ∧ top(p)≠c
      ∧ at(p, d) ∧ at(p', d) ∧ p' ≠ p
      ∧ loc(r)=d ∧ cargo(r)=nil
  body: while top(p) ≠ c do
        c' ← top(p)
        load(r, c', pos(c'), p, d)
        unload(r, c', top(p'), p', d)
  
```

```

r1, c1, p1, d1, p2, d2
m2-put-in-pile(r, c, p, d, p', d')
  task: put-in-pile(c, p')
  pre: pile(c)=p ∧ at(p, d) ∧ at(p', d')
      ∧ p ≠ p' ∧ cargo(r)=nil
  body: if loc(r) ≠ d then navigate(r, d)
        uncover(c)
        load(r, c, pos(c), p, d)
        if loc(r) ≠ d' then navigate(r, d')
        unload(r, c, top(p'), p', d)
  
```

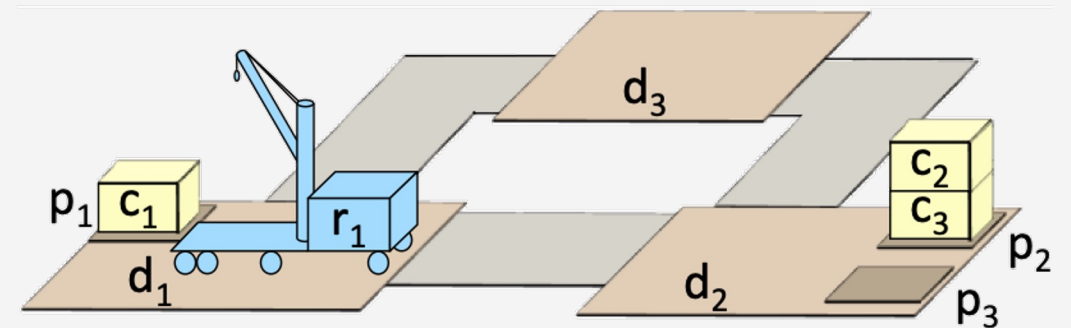


# Example

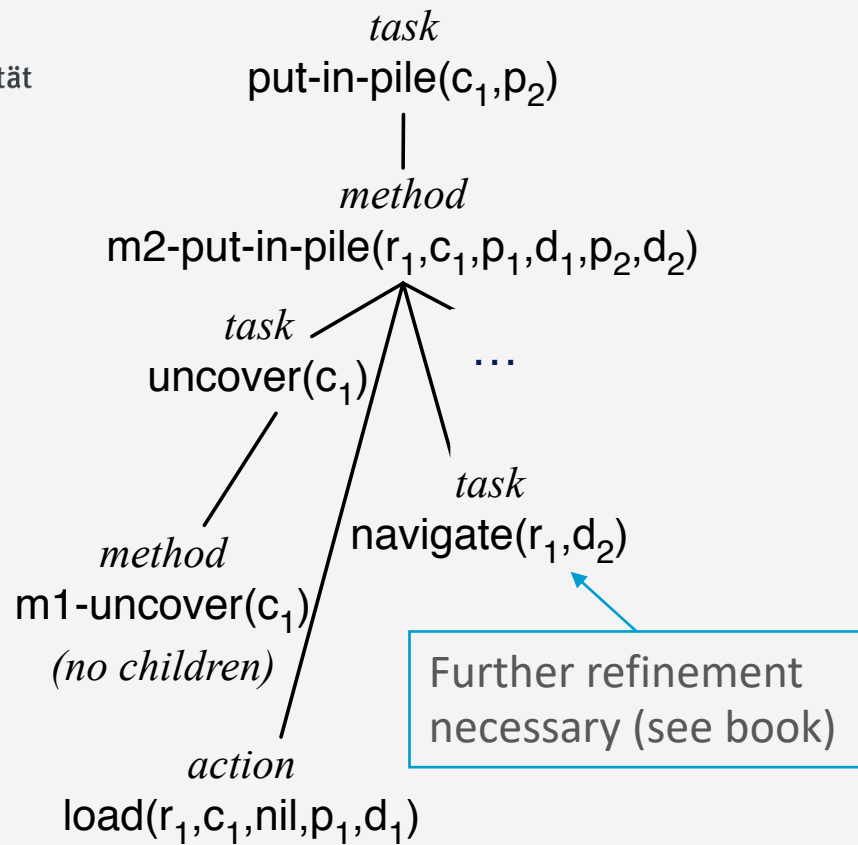


```

    r1,c1,p1,d1,p2,d2
    m2-put-in-pile(r, c, p, d, p', d')
    ...
    body: if loc(r) ≠ d then navigate(r,d)
          uncover(c)
          load(r, c, pos(c), p, d) action
          if loc(r) ≠ d' then navigate(r, d')
          unload(r, c, top(p'), p', d)
  
```

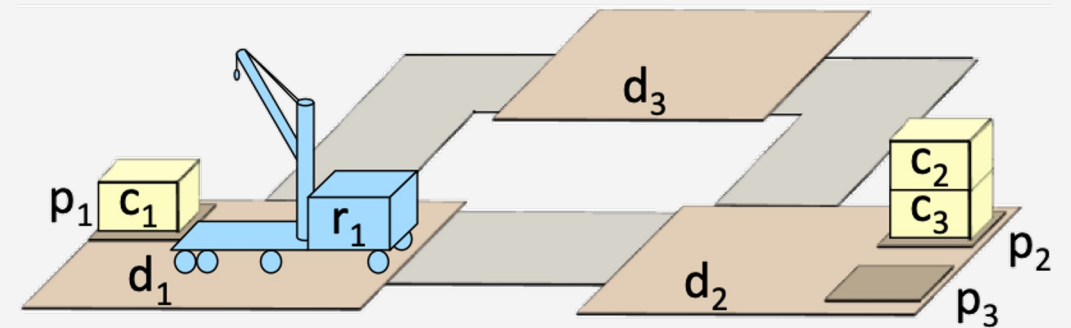


# Example

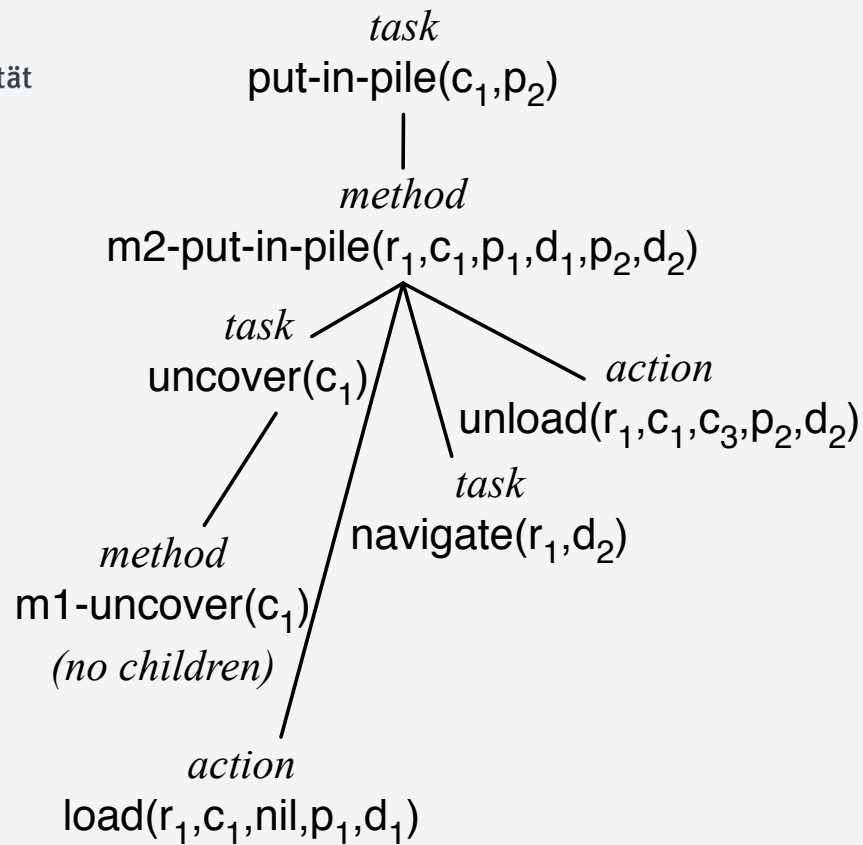


```

    r1,c1,p1,d1,p2,d2
    m2-put-in-pile(r, c, p, d, p', d')
    ...
    body: if loc(r) ≠ d then navigate(r, d)
          uncover(c)
          load(r, c, pos(c), p, d) task
          if loc(r) ≠ d' then navigate(r, d')
          unload(r, c, top(p'), p', d)
  
```

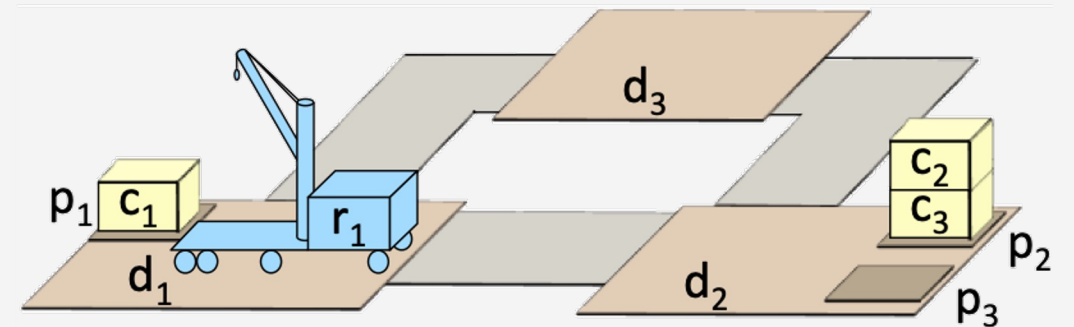


# Example



```

    r1,c1,p1,d1,p2,d2
    m2-put-in-pile(r, c, p, d, p', d')
    ...
    body: if loc(r) ≠ d then navigate(r, d)
          uncover(c)
          load(r, c, pos(c), p, d)
          if loc(r) ≠ d' then navigate(r, d')
          unload(r, c, top(p'), p', d) action
  
```



## Heuristics For SeRPE

- *Ad hoc* approaches:
  - Domain-specific estimates
  - Statistical data on how well each method works
  - Try methods (or actions) in the order that they appear in  $\mathcal{M}$  (or  $\mathcal{A}$ )
- Ideally, would want to implement using heuristic search (e.g., GBFS)
  - What heuristic function? Open problem
- SeRPE is a generalisation of HTN planning
  - In some cases, classical-planning heuristics can be used, in other cases they become intractable [Shivashankar *et al.*, ECAI-2016]

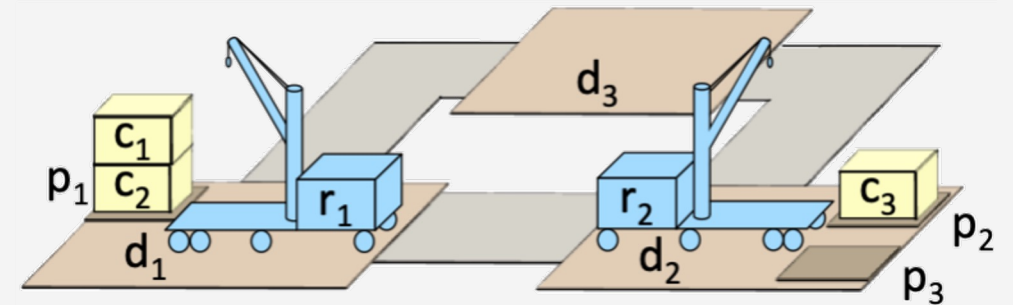
```
SeRPE( $\mathcal{M}, \mathcal{A}, s, \tau$ )  
  Candidates  $\leftarrow$  Instances( $\mathcal{M}, \tau, s$ )  
  if Candidates =  $\emptyset$  then  
    return failure  
  nondeterministically choose  $m \in$  Candidates  
  return Progress-to-finish( $\mathcal{M}, \mathcal{A}, s, \tau, m$ )
```



## Interleaving

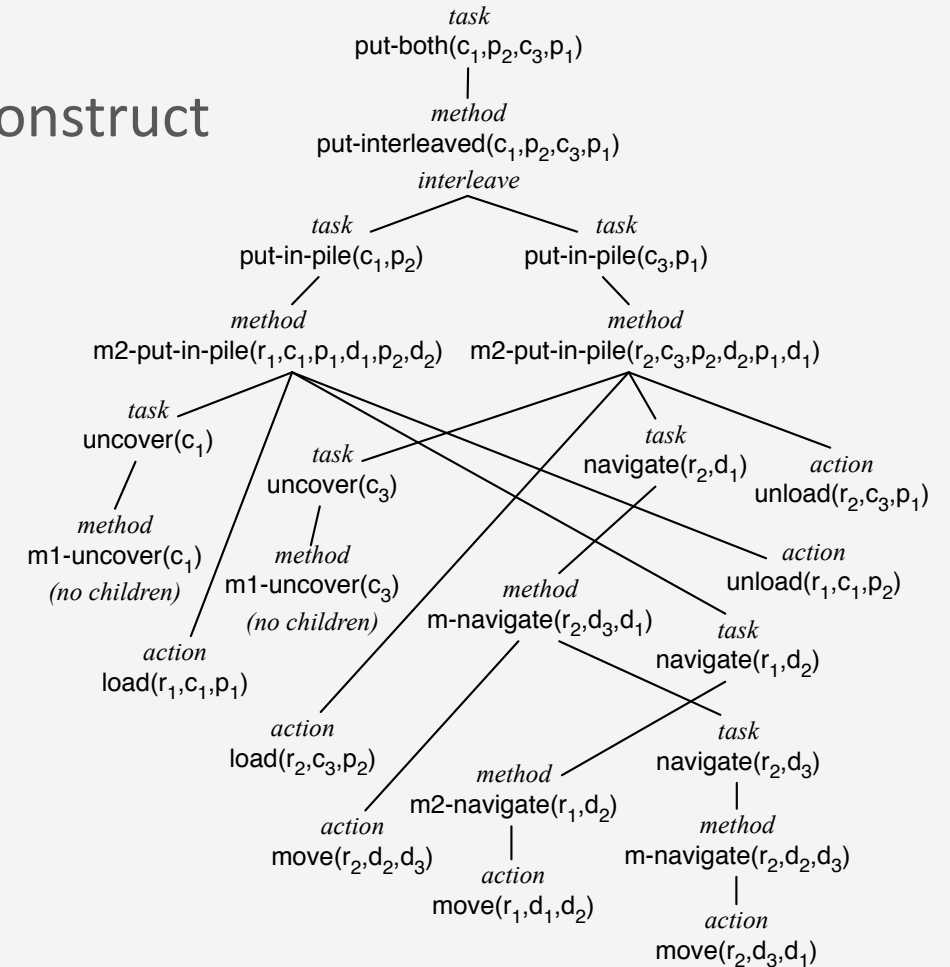
- Want to move  $c_1$  to  $p_2$ , using this plan
  - $\langle \text{load}(r_1, c_1, c_2, p_1, d_1),$   
 $\text{move}(r_1, d_1, d_2),$   
 $\text{unload}(r_1, c_1, p_3, \text{nil}, d_2) \rangle$
- ... and move  $c_3$  to  $p_1$  using this plan:
  - $\langle \text{load}(r_2, c_3, \text{nil}, p_2, d_2),$   
 $\text{move}(r_2, d_2, d_3),$   
 $\text{move}(r_2, d_3, d_1),$   
 $\text{unload}(r_2, c_3, c_2, p_1, d_1) \rangle$

- For it to work, must interleave the plans
  - $\langle \text{load}(r_2, c_3, \text{nil}, p_2, d_2),$   
 $\text{move}(r_2, d_2, d_3),$   
 $\text{load}(r_1, c_1, c_2, p_1, d_1),$   
 $\text{move}(r_1, d_1, d_2),$   
 $\text{unload}(r_1, c_1, p_3, \text{nil}, d_2),$   
 $\text{move}(r_2, d_3, d_1),$   
 $\text{unload}(r_2, c_3, c_2, p_1, d_1) \rangle$



## Interleaved Refinement Tree (IRT) Procedure

- SeRPE does not allow the “concurrent” programming construct
- Partial fix:  
extend SeRPE to interleave plans for different tasks
- Details:  
Section 3.3.2



## Descriptive Action Models

- Predict the outcome of performing a command
  - Preconditions-and-effects representation
- Command
  - $take(r, o, l)$ : robot  $r$  takes object  $o$  at location  $l$
  - $put(r, o, l)$ :  $r$  puts  $o$  at location  $l$
  - $perceive(r, l)$ : robot  $r$  perceives what objects are at location  $l$ 
    - Can only perceive what is at its current location
    - If we knew this in advance, perception would not be necessary

- Action model

$take(r, o, l)$

pre:  $cargo(r) = nil, loc(r) = l, loc(o) = l$

eff:  $cargo(r) \leftarrow o, loc(o) \leftarrow r$

$put(r, o, l)$

pre:  $loc(r) = l, loc(o) = r$

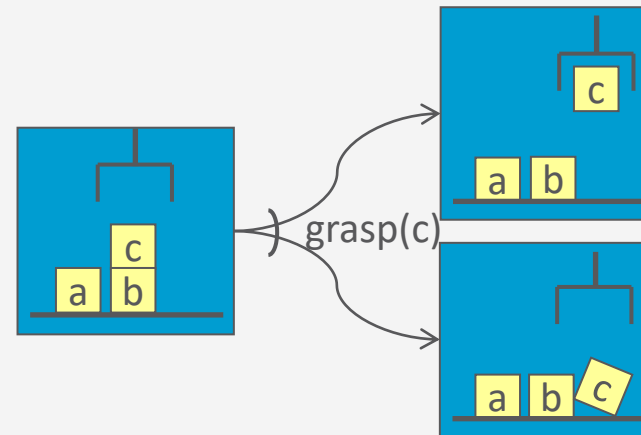
eff:  $cargo(r) \leftarrow nil, loc(o) \leftarrow l$

$perceive(r, l)$

?

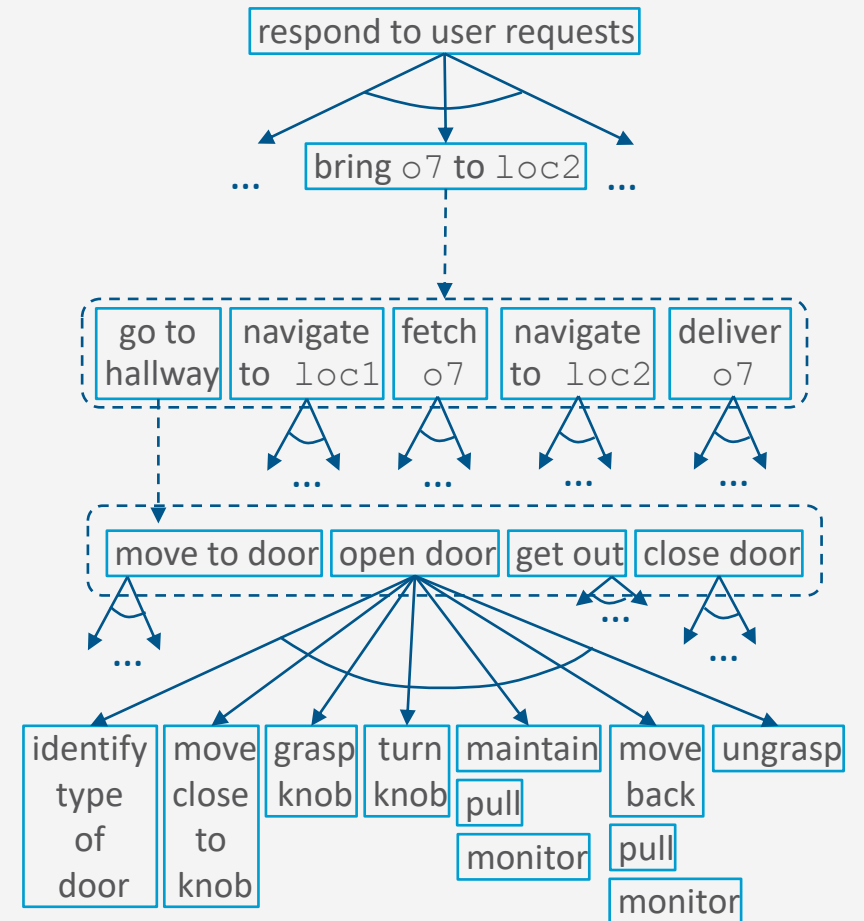
## Limitation

- Most environments are inherently nondeterministic
  - Deterministic action models will not always make the right prediction
- Why use them?
  - Deterministic models  $\Rightarrow$  much simpler planning algorithms
    - Use when errors are infrequent and do not have severe consequences
    - Actor can fix the errors online



# Planning/Acting at Different Levels

- Deterministic models may work better at some levels than others
- May want
  - RAE at some levels
  - RAE+planner at some levels
  - Planner at some levels
- In some cases, might want the planner to reason about nondeterministic outcomes
  - Later in lecture (Book: Ch. 5 + 6)
- Ongoing research on extending refinement planning to handle nondeterminism [Patra *et al.*, AAI-2019]



Planning

Acting

## Summary

- Refinement planning (SeRPE)
  - Plan by simulating RAE on a single external task/event/goal
  - Deterministic actions
    - OK if we are confident of outcome, can recover if things go wrong
  - Interleaved plans (brief example)

## Outline per the Book

### 3.1 *Representation*

- State variables, commands, refinement methods
- Example

### 3.2 *Acting*

- RAE (Refinement Acting Engine)
- Example
- Extensions

### 3.3 *Planning*

- Motivation and basic ideas
- Deterministic action models
- SeRPE (Sequential Refinement Planning Engine)

### **3.4 *Using Planning in Acting***

- Techniques
- Caveats

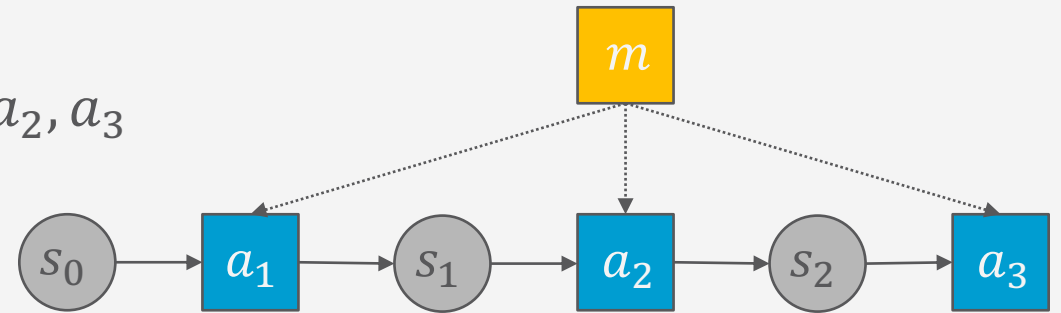
## Acting and Refinement Planning

- Hierarchical acting with refinement planning
  - REAP: a RAE-like actor uses SeRPE-like planning at all levels
- Non-hierarchical actor with refinement planning
  - Refine-Lookahead, Refine-Lazy-Lookahead, Refine-Concurrent-Lookahead
    - Essentially the same as
      - Run-Lookahead, Run-Lazy-Lookahead, Run-Concurrent-Lookahead
  - But they call SeRPE instead of a classical planner for acting out a given task
  - Lookahead same as before
    - Receding horizon, sampling, subgoaling



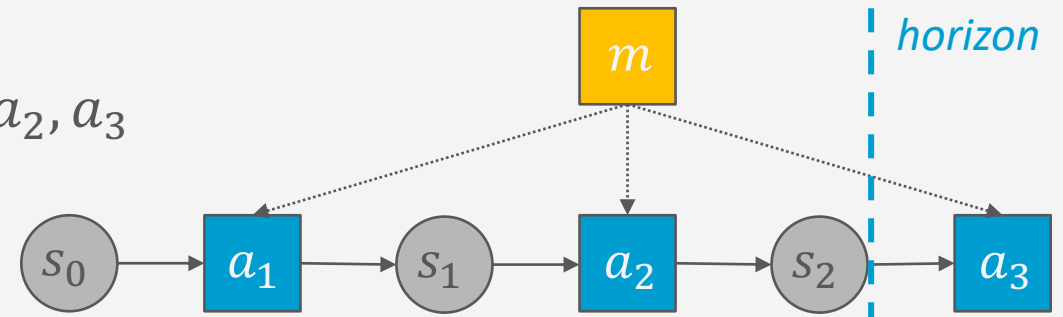
## Caveats

- Start in state  $s_0$ , want to accomplish task  $\tau$ 
  - Refinement method  $m$  with task  $\tau$ , pre  $s_0$ , body  $a_1, a_2, a_3$
- Actor uses Refine-Lookahead
  - Lookahead = SeRPE, returns  $\langle a_1, a_2, a_3 \rangle$
  - Actor performs  $a_1$ , calls Lookahead again
  - No applicable method for  $\tau$  in  $s_1$ , SeRPE returns failure
- Fixes
  - When writing refinement methods, make them general enough to work in different states
  - In some cases, Lookahead might be able to fall back on classical planning until it finds something that matches a method
  - Keep snapshot of SeRPE's search tree at  $s_1$ , resume there next time



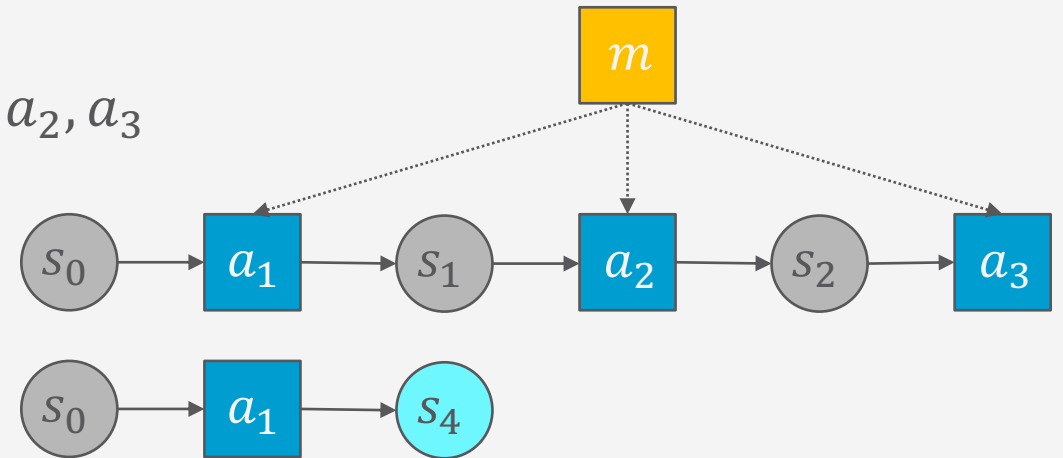
## Caveats

- Start in state  $s_0$ , want to accomplish task  $\tau$ 
  - Refinement method  $m$  with task  $\tau$ , pre  $s_0$ , body  $a_1, a_2, a_3$
- Actor uses Refine-Lazy-Lookahead
  - Lookahead = SeRPE with receding horizon, returns  $\langle a_1, a_2 \rangle$
  - Actor performs them, calls Lookahead again
  - No applicable method for  $\tau$  in  $s_2$ , SeRPE returns failure
- Fixes
  - Can use the same fixes on previous slide, with one modification
    - Keep snapshot of SeRPE's search tree **at the horizon**, resume next time it is called



## Caveats

- Start in state  $s_0$ , want to accomplish task  $\tau$ 
  - Refinement method  $m$  with task  $\tau$ , pre:  $s_0$ , body  $a_1, a_2, a_3$
- Actor uses Refine-Lazy-Lookahead
  - Lookahead = SeRPE, returns  $\langle a_1, a_2, a_3 \rangle$
  - While acting, unexpected event
  - Actor calls Lookahead again
  - No applicable method for  $\tau$  in  $s_4$ , SeRPE returns failure
- Fixes
  - Can use most of the fixes on last two slides, with this modification
    - Keep snapshot of SeRPE's search tree **after each action**
      - In example: restart it immediately after  $a_1$ , using  $s_4$  as current state
    - Also: make **recovery methods** for unexpected states



## Summary

- Acting and planning
  - Lookahead: search part of the search space, return a partial solution
  - Refine-Lookahead, Refine-Lazy-Lookahead, Refine-Concurrent-Lookahead
    - Like Run-Lookahead, Run-Lazy-Lookahead, Run-Concurrent-Lookahead, but call SeRPE
  - Caveats
    - Current state may not be what we expect
    - Possible ways to handle that

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### 3.3 *Planning*

- Motivation and basic ideas
- Deterministic action models
- SeRPE (Sequential Refinement Planning Engine)

### 3.4 *Using Planning in Acting*

- Techniques
- Caveats

⇒ Next: Planning and Acting with Temporal Models