

Perception

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Representing Knowledge

Reminder: The basic agent algorithm

```
1 W is knowledge base
2 g is an unsatisfied goal in W
3 B set of actions available
4 P set of percepts available
5
6 while g not satisfied:
7     PERCEIVE() using percepts in P to update W
8     CHOOSE() action b (from B) that advances towards g
9     EXECUTE() action b
```

- ▶ Knowledge base W starts with initial knowledge \Leftarrow **the focus**
- ▶ PERCEIVE() adds knowledge

Knowledge Representation

- ▶ Whole area of AI devoted to knowledge representation
- ▶ Commonly known as **KR**
- ▶ Own conferences
 - ▶ KRR, knowledge representation and reasoning
- ▶ Lots of thought on how to reason about knowledge
 - ▶ Look up *ontology*, *description logics*, etc.

Simple KR used *here* and *now*

- ▶ Keep track of *beliefs*
 - ▶ We will discuss procedures later
- ▶ Each belief: tuple $\langle k, v \rangle$
 - ▶ k is key (can have variable parameters)
 - ▶ v is a value

Fluents

- ▶ In many areas of AI, logic is dominant representation approach
- ▶ *Fluents* used as representation (non-procedural knowledge)
- ▶ fluent $f = v$:
 - ▶ belief with key f has value v
- ▶ This looks suspiciously like predicates. . . But no.
 - ▶ fluents can change values

Fluent name and parameters

- ▶ A fluent $f(e_1, \dots, e_n)$:
 - ▶ f is fluent name
 - ▶ n is the fluent arity (including 0, $f()$)
 - ▶ e_i entities in a known set E_i , or variables
- ▶ Ground fluents: no variables, e_i constants

Fluent values

- ▶ Fluents can have values from a range V_f
- ▶ *fluent literal*: $f(\dots) = v$, where $v \in V_f$
- ▶ Value can be simple (true/false, number)
- ▶ Value can be complex (data structure)

Examples

- ▶ `robot_at_cell_x() = true`

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- ▶ $\text{robot_at}(x) = \text{true}$
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- ▶ $\text{at}(r,x) = \text{true}$

Examples

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- ▶ $\text{robot_at}(x) = \text{true}$
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- ▶ $\text{at}(r,x) = \text{true}$
- ▶ $\text{at}(r) = x$

Using fluents to represent environment states

- ▶ Given set of Fluents F
- ▶ Environment: collection of states
- ▶ State s is set of fluent literals
 - ▶ Each assigned a value (i.e., is *literal*)
 - ▶ *Consistent*: no $f = v_1, f = v_2$, s.t. $f \in F, v_1 \neq v_2$
 - ▶ *Maximal*: $\forall f \in F, (f = v) \in s$, i.e., $|s| = |F|$
- ▶ Partial state – same def, but not maximal
 - ▶ Really, a collection of states that share fluent literals.
- ▶ This is a **factored state representation**
 - ▶ State is composed of multiple factors

Perception in the Agent

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- ▶ Knowledge base W : collection of fluents
- ▶ PERCEIVE() adds knowledge \Leftarrow **the focus**

Perception

Naive view

- ▶ Update the knowledge-base available to the agent
 - ▶ i.e., Transform environment features to knowledge-base
- ▶ Information from sensors *overwrites* knowledge-base
- ▶ Examples:
 - ▶ Current range straight ahead
 - ▶ Current time
 - ▶ Current location
 - ▶ Current map
 - ▶ Current objects seen

Perception

Naive view

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What's so complicated?

Perception computationally complex

- ▶ Extracting information from current readings
 - ▶ e.g., “Current objects seen” requires computer vision
- ▶ May also require multiple readings over time
 - ▶ “Current map”, “current location” requires SLAM
 - ▶ Expensive, and requires memory

Perception conceptually complex: Body

- ▶ Requires interpretation (sensors never perfect)
 - ▶ At best, have a model of the sensor
- ▶ Perception takes time (e.g., for computation)
- ▶ Integrating information from multiple sensors
 - ▶ **Sensor Fusion**: whole field!
 - ▶ e.g., “obstacle near” if one of many sensors active

Will come back to this later.

Perception conceptually complex: Environment

- ▶ *Transparent* environment: everything perceivable.
 - ▶ May require **attention process**
- ▶ *Translucent* environment: Not everything perceivable
 - ▶ May require **active perception**

Will come back to this later.

Impact of embodiment on agent algorithm

Perception conceptually complex: Body

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Integrating and interpreting information from sensors

- ▶ From multiple sensors
- ▶ From multiple readings (over time)
- ▶ Conceptually PERCEIVE() into (at least) two processes:
 - ▶ $K \leftarrow UPDATE(W)$ which generates new beliefs K (set)
 - ▶ $W_{new} \leftarrow REVISE(K, W)$ which integrates new beliefs with old
- ▶ Separates the act of sensing from the integration over time
- ▶ Reasoning using sensor models, etc.: in REVISE()

Synchronous perception (Original)

- 1 W is knowledge base
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 - 6 while g not satisfied:
 - 7 PERCEIVE() using percepts in P to update W
 - 8 CHOOSE() action b (from B) that advances towards g
 - 9 EXECUTE() action b
- ▶ Single perception step after single action step
 - ▶ OK if **actuation, perception are instantaneous**
 - ▶ What happens when they are not?

Semi-synchronous algorithm

```
1  W is knowledge base
2  g is an unsatisfied goal in W
3  B set of actions available
4  P set of percepts available
5
6  while g not satisfied:
7      CHOOSE() action b (from B) that advances towards g
8      In parallel:
9          EXECUTE() action b
10         PERCEIVE() using percepts in P to update W
```

- ▶ Perception step in parallel to action step
- ▶ Now action, perception can have duration
- ▶ Still needs both to finish before can continue

Asynchronous algorithm

```
1  W is knowledge base
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4  P set of percepts available
5
6  In parallel:
7      while g not satisfied:
8          CHOOSE() action b (from B) that advances towards g
9          EXECUTE() action b
10     while (true) PERCEIVE() using percepts in P to update W
```

- ▶ *Perception process in parallel to action selection process*
- ▶ Completely independent?
- ▶ **No**, will need write-locks accessing *W*

Impact of environment on agent algorithm

Perception conceptually complex: Environment

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Attention, Active Perception

- ▶ How will these algorithms change to allow:
 - ▶ attention? (deciding which percepts to update)
 - ▶ active perception? (choosing actions that will enable perception?)
- ▶ Attention takes place at the UPDATE() stage
 - ▶ UPDATE(W) takes knowledge into account. But how?
- ▶ Active Perception is even bigger challenge:
 - ▶ Take actions to generate new sensing possibilities
 - ▶ Takes place in CHOOSE()!

Open areas of research!