COMP310
Multi-Agent Systems
Chapter 4 - Practical Reasoning Agents

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Pro-Active Behaviour

• Previously we looked at:
  • Characteristics of an Agent and its Environment
  • The Intentional Stance
  • Translating the Formal Agent model into a Deductive Logic framework

• We said:
  • An intelligent agent is a computer system capable of flexible autonomous action in some environment.
  • Where by flexible, we mean:
    • reactive;
    • pro-active;
    • social.

• This is where we deal with the “proactive” bit, showing how we can program agents to have goal-directed behaviour.
What is Practical Reasoning?

• Practical reasoning is reasoning directed towards actions — the process of figuring out what to do:

  “... Practical reasoning is a matter of weighing conflicting considerations for and against competing options, where the relevant considerations are provided by what the agent desires/values/cares about and what the agent believes...” (Bratman)

• Distinguish practical reasoning from theoretical reasoning.

• Theoretical reasoning is directed towards beliefs.
The components of Practical Reasoning

• Human practical reasoning consists of two activities:
  • *deliberation*:
    • deciding *what* state of affairs we want to achieve
    • the outputs of deliberation are *intentions*;
  • *means-ends reasoning*:
    • deciding *how* to achieve these states of affairs
    • the outputs of means-ends reasoning are *plans*.

• Intentions are a key part of this.
  • The interplay between *beliefs*, *desires* and *intentions* defines how the model works
Intentions in Practical Reasoning

1. Intentions pose problems for agents, who need to determine ways of achieving them.

   If I have an intention to φ, you would expect me to devote resources to deciding how to bring about φ.

2. Intentions provide a “filter” for adopting other intentions, which must not conflict.

   If I have an intention to φ, you would not expect me to adopt an intention ψ that was incompatible with φ.

3. Agents track the success of their intentions, and are inclined to try again if their attempts fail.

   If an agent’s first attempt to achieve φ fails, then all other things being equal, it will try an alternative plan to achieve φ.
4. Agents believe their intentions are possible.

That is, they believe there is at least some way that the intentions could be brought about.

5. Agents do not believe they will not bring about their intentions.

It would not be rational of me to adopt an intention to $\varphi$ if I believed I would fail with $\varphi$.

6. Under certain circumstances, agents believe they will bring about their intentions.

If I intend $\varphi$, then I believe that under “normal circumstances” I will succeed with $\varphi$. 
Intentions in Practical Reasoning

7. Agents need not intend all the expected side effects of their intentions.

- Intentions are not closed under implication.
- This last problem is known as the side effect or package deal problem.

If I believe $\phi \Rightarrow \psi$ and I intend that $\phi$, I do not necessarily intend $\psi$ also.

I may believe that going to the dentist involves pain, and I may also intend to go to the dentist — but this does not imply that I intend to suffer pain!
Intentions are Stronger than Desire

“... My desire to play basketball this afternoon is merely a potential influencer of my conduct this afternoon. It must vie with my other relevant desires [. . . ] before it is settled what I will do.

In contrast, once I intend to play basketball this afternoon, the matter is settled: I normally need not continue to weigh the pros and cons.

When the afternoon arrives, I will normally just proceed to execute my intentions...”

Michael E. Bratman (1990)
Means-ends Reasoning/Planning

- Planning is the design of a course of action that will achieve some desired goal.
  - Basic idea is to give a planning system:
    - (representation of) goal/intention to achieve;
    - (representation of) actions it can perform;
    - (representation of) the environment;
  - and have it generate a plan to achieve the goal.

- This is automatic programming.
Means-ends Reasoning/Planning

- Don't have to directly tell the system what to do!
- Let it **figure out** how to achieve the goal on its own!
STRIPS Planner

• STRIPS
  • The Stanford Research Institute Problem Solver
  • Used by Shakey, the robot
    • Developed by Richard Fikes and Nils Nilsson in 1971 at SRI International
Representations

• **Question**: How do we represent . . .
  • goal to be achieved;
  • state of environment;
  • actions available to agent;
  • plan itself.

• **Answer**: We use *logic*, or something that looks a lot like logic.
We’ll illustrate the techniques with reference to the blocks world.

A simple (toy) world, in this case one where we consider toys

The blocks world contains a robot arm, 3 blocks (A, B and C) of equal size, and a table-top.

The aim is to generate a plan for the robot arm to build towers out of blocks.
Blocksworld

• The environment is represented by an ontology.

• The closed world assumption is used
  • Anything not stated is assumed to be false.

• A goal is represented as a set of formulae.

Blocksworld Ontology
- \(\text{On}(x,y)\) object \(x\) on top of object \(y\)
- \(\text{OnTable}(x)\) object \(x\) is on the table
- \(\text{Clear}(x)\) nothing is on top of object \(x\)
- \(\text{Holding}(x)\) arm is holding \(x\)

Representation of the following blocks
- \(\text{Clear}(A)\)
- \(\text{On}(A, B)\)
- \(\text{OnTable}(B)\)
- \(\text{Clear}(C)\)
- \(\text{OnTable}(C)\)
- \(\text{ArmEmpty}\)

The goal:
\(\{\text{OnTable}(A), \text{OnTable}(B), \text{OnTable}(C), \text{ArmEmpty}\}\)
Blocksworld Actions

• Each action has:
  • a name: which may have arguments;
  • a pre-condition list: list of facts which must be true for action to be executed;
  • a delete list: list of facts that are no longer true after action is performed;
  • an add list: list of facts made true by executing the action.

• Each of these may contain variables.

• What is a plan?
  • A sequence (list) of actions, with variables replaced by constants.
## Blocksworld Actions

**Stack** \((x, y)\)

- **pre**: \(\text{Clear}(y) \land \text{Holding}(x)\)
- **del**: \(\text{Clear}(y) \land \text{Holding}(x)\)
- **add**: \(\text{ArmEmpty} \land \text{On}(x, y)\)

The **stack** action occurs when the robot arm places the object \(x\) it is holding is placed on top of object \(y\).

**Pickup** \((x)\)

- **pre**: \(\text{Clear}(x) \land \text{OnTable}(x) \land \text{ArmEmpty}\)
- **del**: \(\text{OnTable}(x) \land \text{ArmEmpty}\)
- **add**: \(\text{Holding}(x)\)

The **pickup** action occurs when the arm picks up an object \(x\) from the table.

**UnStack** \((x, y)\)

- **pre**: \(\text{On}(x, y) \land \text{Clear}(x) \land \text{ArmEmpty}\)
- **del**: \(\text{On}(x, y) \land \text{ArmEmpty}\)
- **add**: \(\text{Holding}(x) \land \text{Clear}(y)\)

The **unstack** action occurs when the robot arm picks an object \(y\) up from on top of another object \(y\).

**PutDown** \((x)\)

- **pre**: \(\text{Holding}(x)\)
- **del**: \(\text{Holding}(x)\)
- **add**: \(\text{OnTable}(x) \land \text{ArmEmpty} \land \text{Clear}(x)\)

The **putdown** action occurs when the arm places the object \(x\) onto the table.
Using Plans

To get from here (left) to here (right)...

...we need this set of actions:

- \textit{UnStack}(A,B)
- \textit{Putdown}(A)
- \textit{Pickup}(B)
- \textit{Stack}(B, C)
- \textit{Pickup}(A)
- \textit{Stack}(A, B)

<table>
<thead>
<tr>
<th>Stack(x, y))</th>
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</thead>
<tbody>
<tr>
<td>pre \ Clear(y) &amp; \ Holding(x)</td>
</tr>
<tr>
<td>del \ Clear(y) &amp; \ Holding(x)</td>
</tr>
<tr>
<td>add \ ArmEmpty &amp; \ On(x, y)</td>
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</tbody>
</table>

<table>
<thead>
<tr>
<th>UnStack(x, y))</th>
</tr>
</thead>
<tbody>
<tr>
<td>pre \ On(x, y) &amp; \ Clear(x) &amp; \ ArmEmpty</td>
</tr>
<tr>
<td>del \ On(x, y) &amp; \ ArmEmpty</td>
</tr>
<tr>
<td>add \ Holding(x) &amp; \ Clear(y)</td>
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</tbody>
</table>

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<thead>
<tr>
<th>Pickup(x))</th>
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<tr>
<td>pre \ Clear(x) &amp; \ OnTable(x) &amp; \ ArmEmpty</td>
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<tr>
<td>del \ OnTable(x) &amp; \ ArmEmpty</td>
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<tr>
<td>add \ Holding(x)</td>
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</tbody>
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<tr>
<th>PutDown(x))</th>
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<tbody>
<tr>
<td>pre \ Holding(x)</td>
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<tr>
<td>del \ Holding(x)</td>
</tr>
<tr>
<td>add \ OnTable(x) &amp; \ ArmEmpty &amp; \ Clear(x)</td>
</tr>
</tbody>
</table>
Plan Validity

• Thus, a plan is simply a sequence of steps

• However, how can we:
  • Generate the plan?
  • Ensure that it is correct?
Formal Representation

• Let’s relate the STRIPS model back to the formal description of an agent we talked about before.
  • This will help us to see how it fits into the overall picture.

• As before we assume that the agent has a set of actions \( Ac \), and we will write individual actions as \( \alpha_1, \alpha_2 \) and so on.

• Now the actions have some structure, each one has preconditions \( P_{\alpha_i} \), add list \( A_{\alpha_i} \), and delete list \( D_{\alpha_i} \), for each \( \alpha_i \in Ac \):

\[
\alpha_i = (P_{\alpha_i}, D_{\alpha_i}, A_{\alpha_i})
\]

• A plan is just a sequence of actions, where each action is one of the actions from \( Ac \):

\[
\pi = (\alpha_1, \ldots, \alpha_n)
\]
Formal Representation

• A **planning problem** is therefore: \( \langle B_0, Ac, I \rangle \)
  
  - \( B_0 \) is the set of beliefs the agent has about the world.
  - \( Ac \) is the set of actions, and
  - \( I \) is a goal (or intention)

• Since actions change the world, any rational agent will change its beliefs about the world as a result of carrying out actions.
  
  - Thus, a plan \( \pi \) for a given planning problem will be associated with a sequence of sets of beliefs:
    \[
    B_0 \xrightarrow{\alpha_1} B_1 \xrightarrow{\alpha_2} \cdots \xrightarrow{\alpha_n} B_n
    \]
  
  - In other words at each step of the plan the beliefs are updated by removing the items in the delete list of the relevant action and adding the items in the add list.
Formal Representation

• A plan $\pi$ is said to be acceptable with respect to the problem $\langle B_0, Ac, I \rangle$ if and only if, for all $1 \leq j \leq n$, $B_{j-1} \models P_{oj}$

  • In other words, the pre-requisites for each action have to be true right before the action is carried out.
  • We say this because the pre-conditions don’t have to be in $B_{j-1}$, we just have to be able to prove the pre-conditions from $B_{j-1}$.

• A plan $\pi$ is correct if it is acceptable, and: $B_n \models i$

  • In other words, it is correct if it is acceptable and the final state makes the goal true.
Example - Question 2.b

(b) The Blocksworld scenario is represented by an ontology with the following formulae:

- `On(x, y)` object z on top of obj y
- `OnTable(z)` object z is on the table
- `Clear(x)` nothing is on top of obj x
- `Holding(z)` arm is holding x

An agent has a set of actions $A_e$, such that $A_e = \{\text{Stack, UnStack, Pickup, PutDown}\}$:

- **Stack(x, y)**
  - pre `Clear(y)` & `Holding(x)`
  - del `Clear(y)` & `Holding(x)`
  - add `ArmEmpty` & `On(x, y)`

- **UnStack(x, y)**
  - pre `On(x, y)` & `Clear(z)` & `ArmEmpty`
  - del `On(x, y)` & `ArmEmpty`
  - add `Holding(z)` & `Clear(y)`

- **Pickup(x)**
  - pre `Clear(z)` & `OnTable(z)` & `ArmEmpty`
  - del `OnTable(z)` & `ArmEmpty`
  - add `Holding(z)`

- **PutDown(x)**
  - pre `Holding(z)`
  - del `Holding(z)`
  - add `OnTable(z)` & `ArmEmpty` & `Clear(z)`

It also has the following beliefs $B_e$, regarding the three bricks $\{A, B, C\}$, and the intention $i$:

<table>
<thead>
<tr>
<th>Beliefs $B_e$</th>
<th>Intention $i$</th>
</tr>
</thead>
<tbody>
<tr>
<td><code>Clear(B)</code></td>
<td><code>Clear(A)</code></td>
</tr>
<tr>
<td><code>Clear(C)</code></td>
<td><code>Clear(B)</code></td>
</tr>
<tr>
<td><code>On(C, A)</code></td>
<td><code>On(B, C)</code></td>
</tr>
<tr>
<td><code>OnTable(A)</code></td>
<td><code>OnTable(A)</code></td>
</tr>
<tr>
<td><code>OnTable(B)</code></td>
<td><code>OnTable(C)</code></td>
</tr>
<tr>
<td><code>ArmEmpty</code></td>
<td><code>ArmEmpty</code></td>
</tr>
</tbody>
</table>

Calculate a plan $\pi$ that would achieve $i$, given the beliefs $B_e$. Draw the environment at the beginning of the plan, and after every time a `Stack` or `UnStack` action is performed.

(10 marks)
A plan $\pi$ is a sequence of actions, where each action results in changing the set of beliefs the agent has, until the final set of beliefs matches that of the intentions, such that $B_0 \xrightarrow{a_1} B_1 \xrightarrow{a_2} \cdots \xrightarrow{a_n} B_n$. Therefore, a planner will explore all the different possible sequences of actions to determine which one will result in the final set of intentions.

In this solution, only those actions that result in the final solution are given, with the set of beliefs that result in each step presented. The aim is to start with an initial set of beliefs, $B_0$, and arrive at a final set of beliefs, $B_n$ which corresponds to the intentions given in the question - i.e.

<table>
<thead>
<tr>
<th>Beliefs $B_0$</th>
<th>Intention $i$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\text{Clear}(B)$</td>
<td>$\text{Clear}(A)$</td>
</tr>
<tr>
<td>$\text{Clear}(C)$</td>
<td>$\text{Clear}(B)$</td>
</tr>
<tr>
<td>$\text{On}(C, A)$</td>
<td>$\text{On}(B, C)$</td>
</tr>
<tr>
<td>$\text{OnTable}(A)$</td>
<td>$\text{OnTable}(A)$</td>
</tr>
<tr>
<td>$\text{OnTable}(B)$</td>
<td>$\text{OnTable}(C)$</td>
</tr>
<tr>
<td>$\text{ArmEmpty}$</td>
<td>$\text{ArmEmpty}$</td>
</tr>
</tbody>
</table>

The solution is given on the next slide. In each case, the beliefs that hold prior to the action are given in bold, and the beliefs that are new after the action are also presented in bold.
The beliefs B4, once rearranged, are now equivalent to the intentions.
Action Definitions are important!!!

One of these definitions works. The other doesn't!
Implementing Practical Reasoning Agents

• A first pass at an implementation of a practical reasoning agent:

• For now we will not be concerned with stages 2 or 3.

• These are related to the functions see and next from the earlier lecture notes.

Agent Control Loop Version 1
1. while true
2. observe the world;
3. update internal world model;
4. deliberate about what intention to achieve next;
5. use means-ends reasoning to get a plan for the intention;
6. execute the plan
7. end while
Implementing Practical Reasoning Agents

• see is as before:
  • see: $E \rightarrow \text{Percept}$

• Instead of the function next...
  • which took a percept and used it to update the internal state of an agent

• ...we have a belief revision function:
  • $\text{brf}: \mathcal{P}\{\text{Bel}\} \times \text{Percept} \rightarrow \mathcal{P}\{\text{Bel}\}$
    • $\mathcal{P}\{\text{Bel}\}$ is the power set of beliefs
    • Bel is the set of all possible beliefs that an agent might have.
Implementing Practical Reasoning Agents

• **Problem:**
  • deliberation and means-ends reasoning processes are not instantaneous.
  • They have a *time cost*.

• **Suppose that deliberation is **optimal****
  • The agent selects the optimal intention to achieve, then this is the best thing for the agent.
  • i.e. it maximises expected utility.

• So the agent selects an intention to achieve that would have been optimal *at the time it observed the world*.
  • This is calculative rationality.

• The *world may change* in the meantime.
  • Even if the agent can compute the right thing to do, it may not do the right thing.
  • Optimality is hard.
Implementing Practical Reasoning Agents

• Let’s make the algorithm more formal with the algorithm opposite
  • where $I \subseteq \text{Int}$, i.e the set of intentions,
  • $\text{plan}()$ is exactly what we discussed above,
  • $\text{brf}()$ is the belief revision function,
  • and $\text{execute}()$ is a function that executes each action in a plan.

• How might we implement these functions?

Agent Control Loop Version 2
1. $B := B_0$; /* initial beliefs */
2. while true do
3. get next percept $\rho$;
4. $B := \text{brf}(B, \rho)$;
5. $I := \text{deliberate}(B)$;
6. $\pi := \text{plan}(B, I)$;
7. $\text{execute}(\pi)$
8. end while
Deliberation

• How does an agent *deliberate*?
  • begin by trying to understand what the *options* available to you are;
  • *choose* between them, and *commit* to some.

• Chosen options are then *intentions*.

• The *deliberate* function can be decomposed into two distinct functional components:
  • *option generation*; and
  • *filtering*.
Option Generation and Filtering

• Option Generation
  • In which the agent generates a set of possible alternatives
  • Represent option generation via a function, options(), which takes the agent’s current beliefs and current intentions, and from them determines a set of options
    • desires

\[
\text{options} : \mathcal{P} (\text{Bel}) \times \mathcal{P} (\text{Int}) \rightarrow \mathcal{P} (\text{Des})
\]

• Filtering
  • In which the agent chooses between competing alternatives, and commits to achieving them.
  • In order to select between competing options, an agent uses a filter() function.
    • intentions

\[
\text{filter} : \mathcal{P} (\text{Bel}) \times \mathcal{P} (\text{Des}) \times \mathcal{P} (\text{Int}) \rightarrow \mathcal{P} (\text{Int})
\]
Implementing Practical Reasoning Agents

Agent Control Loop Version 3

1. \( B := B_0; \)
2. \( I := I_0; \)
3. while true do
4. get next percept \( \rho; \)
5. \( B := \text{brf}(B, \rho); \)
6. \( D := \text{options}(B, I); \)
7. \( I := \text{filter}(B, D, I); \)
8. \( \pi := \text{plan}(B, I); \)
9. \( \text{execute}(\pi) \)
10. end while
Under Commitment

“... Some time in the not-so-distant future, you are having trouble with your new household robot. You say “Willie, bring me a beer.” The robot replies “OK boss.” Twenty minutes later, you screech “Willie, why didn’t you bring me that beer?” It answers “Well, I intended to get you the beer, but I decided to do something else.” Miffed, you send the wise guy back to the manufacturer, complaining about a lack of commitment...”

“... After retrofitting, Willie is returned, marked “Model C: The Committed Assistant.” Again, you ask Willie to bring you a beer. Again, it accedes, replying “Sure thing.” Then you ask: “What kind of beer did you buy?” It answers: “Genessee.” You say “Never mind.” One minute later, Willie trundles over with a Genessee in its gripper...”
“... After still more tinkering, the manufacturer sends Willie back, promising no more problems with its commitments. So, being a somewhat trusting customer, you accept the rascal back into your household, but as a test, you ask it to bring you your last beer. [. . . ]

The robot gets the beer and starts towards you. As it approaches, it lifts its arm, wheels around, deliberately smashes the bottle, and trundles off. Back at the plant, when interrogated by customer service as to why it had abandoned its commitments, the robot replies that according to its specifications, it kept its commitments as long as required — commitments must be dropped when fulfilled or impossible to achieve. By smashing the bottle, the commitment became unachievable...”

Degrees of Commitment

• Blind commitment
  • A blindly committed agent will continue to maintain an intention until it believes the intention has actually been achieved. Blind commitment is also sometimes referred to as *fanatical* commitment.

• Single-minded commitment
  • A single-minded agent will continue to maintain an intention *until* it believes that *either* the intention has been achieved, *or else that it is no longer possible* to achieve the intention.

• Open-minded commitment
  • An open-minded agent will maintain an intention *as long as it is still believed* possible.
Degrees of Commitment

- An agent has commitment both to:
  - **ends** (i.e., the state of affairs it wishes to bring about), and
  - **means** (i.e., the mechanism via which the agent wishes to achieve the state of affairs).

- Currently, our agent control loop is overcommitted, both to means and ends.
  - Modification: *replan* if ever a plan goes wrong.
  - However, to write the algorithm down we need to refine our notion of plan execution.
Degrees of Commitment

• The previous version was **blindly committed to its means and its ends**

• If \( \pi \) is a plan, then:
  • \( empty(\pi) \) is true if there are no more actions in the plan.
  • \( hd(\pi) \) returns the first action in the plan.
  • \( tail(\pi) \) returns the plan minus the head of the plan.
  • \( sound(\pi, I, B) \) means that \( \pi \) is a correct plan for \( I \) given \( B \).

Agent Control Loop Version 4

1. \( B := B_0; \)
2. \( I := I_0; \)
3. while true do
4.   get next percept \( \rho; \)
5.   \( B := brf(B, \rho); \)
6.   \( D := options(B, I); \)
7.   \( I := filter(B, D, I); \)
8.   \( \pi := plan(B, I); \)
9.   while not empty(\( \pi \)) do
10.   \( \alpha := hd(\pi); \)
11.   execute(\( \alpha \));
12.   \( \pi := tail(\pi); \)
13.   get next percept \( \rho; \)
14.   \( B := brf(B, \rho); \)
15.   if not sound(\( \pi, I, B \)) then
16.     \( \pi := plan(B, I); \)
17.   end-if
18. end-while
19. end-while
Degrees of Commitment

- **Blind Commitment**
  - Makes the control loop more reactive, able to change intention when the world changes.
  - i.e. it is not committed to its means (line 16)

- Still overcommitted to intentions (ends).
  - Never stops to consider whether or not its intentions are appropriate.

---

Agent Control Loop Version 4

1. \( B \) := \( B_0 \);
2. \( I \) := \( I_0 \);
3. while true do
   4. get next percept \( \rho \);
   5. \( B \) := \( brf(B, \rho) \);
   6. \( D \) := \( options(B, I) \);
   7. \( I \) := \( filter(B, D, I) \);
   8. \( \pi \) := \( plan(B, I) \);
   9. while not empty(\( \pi \)) do
      10. \( \alpha \) := \( hd(\pi) \);
      11. execute(\( \alpha \));
      12. \( \pi \) := \( tail(\pi) \);
      13. get next percept \( \rho \);
      14. \( B \) := \( brf(B, \rho) \);
      15. if not sound(\( \pi, I, B \)) then \( \pi \) := \( plan(B, I) \)
      16. end-if
   17. end-while
18. end-while
Single Minded Commitment

• Modification:
  • stop to determine whether intentions have succeeded or whether they are impossible

• Our agent now gets to reconsider its intentions once every time around the outer control loop (line 9), i.e., after:
  • it has completely executed a plan to achieve its current intentions; or
  • it believes it has achieved its current intentions; or
  • it believes its current intentions are no longer possible.
Open Minded Commitment

- Open Minded Commitment
  - In the previous version, our agent reconsiders its intentions once every time around the outer control loop
  - In this new version, our agent also reconsiders its intentions after every action (lines 15 & 16)

- But this intention reconsideration is costly!
Intention Reconsideration

• A dilemma:
  • an agent that does not stop to reconsider its intentions sufficiently often will continue to attempt to achieve its intentions even after it is clear that they cannot be achieved, or that there is no longer any reason for achieving them;
  • an agent that constantly reconsiders its attentions may spend insufficient time actually working to achieve them, and hence runs the risk of never actually achieving them.

• Solution: incorporate an explicit meta-level control component, that decides whether or not to reconsider.
### Intention Reconsideration

- The possible interactions between meta-level control and deliberation are:

<table>
<thead>
<tr>
<th>Situation number</th>
<th>Chose to deliberate?</th>
<th>Changed intentions?</th>
<th>Would have changed intentions?</th>
<th>\texttt{reconsider(...)}</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>No</td>
<td>—</td>
<td>No</td>
<td>Yes</td>
</tr>
<tr>
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<td>No</td>
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<td>No</td>
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<tr>
<td>4</td>
<td>Yes</td>
<td>Yes</td>
<td>—</td>
<td>Yes</td>
</tr>
</tbody>
</table>

- An important assumption: cost of \texttt{reconsider(...)} is much less than the cost of the deliberation process itself.
In situation (1), the agent did not choose to deliberate, and as a consequence, did not choose to change intentions. Moreover, if it had chosen to deliberate, it would not have changed intentions. In this situation, the \texttt{reconsider(...)} function is behaving optimally.

In situation (2), the agent did not choose to deliberate, but if it had done so, it would have changed intentions. In this situation, the \texttt{reconsider(...)} function is not behaving optimally.

In situation (3), the agent chose to deliberate, but did not change intentions. In this situation, the \texttt{reconsider(...)} function is not behaving optimally.

In situation (4), the agent chose to deliberate, and did change intentions. In this situation, the \texttt{reconsider(...)} function is behaving optimally.

\begin{table}[h]
\centering
\begin{tabular}{c|c|c|c|c}
\hline
Situation number & Chose to deliberate? & Changed intentions? & Would have changed intentions? & \texttt{reconsider(...)} optimal? \\
\hline
1 & No & — & No & Yes \\
2 & No & — & Yes & No \\
3 & Yes & No & — & No \\
4 & Yes & Yes & — & Yes \\
\hline
\end{tabular}
\end{table}

\begin{verbatim}
if \texttt{reconsider}(I,B) then
    D := \texttt{options}(B,I);
    I := \texttt{filter}(B,D,I);
end-if
\end{verbatim}
Optimal Intention Reconsideration

• Kinny and Georgeff’s experimentally investigated effectiveness of intention reconsideration strategies.

• Two different types of reconsideration strategy were used:
  • **bold** agents: never pause to reconsider intentions, and
  • **cautious** agents: stop to reconsider after every action.

• **Dynamism** in the environment is represented by the rate of world change, $\gamma$.
  • Experiments were carried out using Tileword.
Optimal Intention Reconsideration

- If $\gamma$ is low (i.e., the environment does not change quickly), then bold agents do well compared to cautious ones.
  - This is because cautious ones waste time reconsidering their commitments while bold agents are busy working towards — and achieving — their intentions.

- If $\gamma$ is high (i.e., the environment changes frequently), then cautious agents can outperform bold agents.
  - This is because they are able to recognise when intentions are doomed, and also to take advantage of serendipitous situations and new opportunities when they arise.

- When planning costs are high, this advantage can be eroded.
Implemented BDI Agents: Procedural Reasoning System

• We now make the discussion even more concrete by introducing an actual agent architecture: the Procedural Reasoning System (PRS).

• In the PRS, each agent is equipped with a **plan library**, representing that agent’s **procedural knowledge**: knowledge about the mechanisms that can be used by the agent in order to realise its intentions.

• The options available to an agent are directly determined by the plans an agent has: an agent with no plans has no options.

• In addition, PRS agents have explicit representations of beliefs, desires, and intentions, as above.
Example PRS (JAM) System

• The agent possesses a number of pre-compiled plans (constructed manually)
  • Each plan contains:
    • a goal - the postcondition of the plan
    • a context - the precondition of the plan
    • a body - the course of action to take out

• When an agent starts, goals are pushed onto the intention stack.
  • This stack contains all of the goals that are pending
  • A set of facts or beliefs are maintained and updated as the agent achieves different goals
  • The agent then deliberates (i.e. selects the most appropriate goal to adopt).
    • This is achieved using meta level plans, or utilities
    • When utilities are used, the agent selects the goal with the highest value

GOALS:
  ACHIEVE blocks_stacked;

FACTS:
  // Block1 on Block2 initially so need //to clear Block2 before stacking.
  FACT ON "Block1" "Block2";
  FACT ON "Block2" "Table";
  FACT ON "Block3" "Table";
  FACT CLEAR "Block1";
  FACT CLEAR "Block3";
  FACT CLEAR "Table";
  FACT initialized "False";
Example PRS (JAM) System

• This is the plan for the top level goal:
  • ACHIEVE blocks_stacked.

• Note that the body contains a mix of instructions and goals.

• When executing, the goals will be added to the intention stack
Example PRS (JAM) System

• This plan also has a utility associated with it
  • This is used by the agent during the deliberation phase

• The plan can also determine actions to execute if it fails

Plan: {
  NAME: "Stack blocks that are already clear"
  GOAL:
    ACHIEVE ON $OBJ1 $OBJ2;
  CONTEXT:
  BODY:
    EXECUTE print "Making sure "$OBJ1" is clear\n";
    ACHIEVE CLEAR $OBJ1;
    EXECUTE print "Making sure "$OBJ2" is clear.\n";
    ACHIEVE CLEAR $OBJ2;
    EXECUTE print "Moving "$OBJ1" on top of "$OBJ2".\n";
    PERFORM move $OBJ1 $OBJ2;
  UTILITY: 10;
  FAILURE:
    EXECUTE print \n"\nStack blocks failed!\n\n";
}
Example PRS (JAM) System

- This plan includes an \textbf{EFFECTS} field

- This determines what the agent should do once the agent has succeeded in executing all of the \textbf{BODY} instructions.

Plan: {
  \textbf{NAME}: "Clear a block"
  \textbf{GOAL}:
  \hspace{1em}ACHIEVE CLEAR $OBJ;
  \textbf{CONTEXT}:
  \hspace{1em}FACT ON $OBJ2 $OBJ;
  \textbf{BODY}:
  \hspace{2em}EXECUTE print "Clear " $OBJ2 " from on top of " $OBJ "\n";
  \hspace{2em}EXECUTE print "Move " $OBJ2 " to table.\n";
  \hspace{2em}ACHIEVE ON $OBJ2 "Table";
  \textbf{EFFECTS}:
  \hspace{3em}EXECUTE print "Clear: Retract ON " $OBJ2 " " $OBJ "\n";
  \hspace{3em}RETRACT ON $OBJ1 $OBJ;
  \textbf{FAILURE}:
  \hspace{4em}EXECUTE print "\n\nClearing block " $OBJ " failed!\n\n";
}
Example PRS (JAM) System

• I’ll leave it as an exercise to work out what plans will be executed
  • If you have a solution and want me to check, let me know.

• The Java version of JAM and further details/documentation are available from Marcus Huber’s website:
  • http://www.marcush.net/IRS/irs_downloads.html

Plan: {
  NAME: "Move a block onto another object"
  GOAL:
    PERFORM move $OBJ1 $OBJ2;
  CONTEXT:
    FACT CLEAR $OBJ1;
    FACT CLEAR $OBJ2;
  BODY:
    EXECUTE print "Performing low-level move action"
    EXECUTE print " of "$OBJ1 " to " $OBJ2 "."
  EFFECTS:
    WHEN : TEST (!= $OBJ2 "Table") {
      EXECUTE print "-Retract CLEAR " $OBJ2 "\n"
      RETRACT CLEAR $OBJ2;
    }
    FACT ON $OBJ1 $OBJ3;
    EXECUTE print "-move: Retract ON "$OBJ1 " "$OBJ3 "\n"
    RETRACT ON $OBJ1 $OBJ3;
    EXECUTE print "-move: Assert CLEAR "$OBJ3 "\n"
    ASSERT CLEAR $OBJ3;
    EXECUTE print "-move: Assert ON "$OBJ1 " "$OBJ2 "\n"
    ASSERT ON $OBJ1 $OBJ2;
  FAILURE:
    EXECUTE print "\n\nMove failed!\n\n";
}
Summary

• This lecture has covered a lot of ground on practical reasoning.
  • We started by discussing what practical reasoning was, and how it relates to intentions.
  • We then looked at planning (how an agent achieves its desires) and how deliberation and means-ends reasoning fit into the basic agent control loop.
  • We then refined the agent control loop, considering commitment and intention reconsideration.
  • Finally, we looked at an implemented system (the textbook discusses a couple of others).

• Next Lecture - we start looking at the AgentSpeak and the Jason framework to exploit BDI.

Class Reading (Chapter 4):

This is an interesting, insightful article, with not too much technical content. It introduces the IRMA architecture for practical reasoning agents, which has been very influential in the design of subsequent systems.