# LECTURE 4: PRACTICAL REASONING

An Introduction to MultiAgent Systems http://www.csc.liv.ac.uk/~mjw/pubs/imas

#### 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)
- Practical reasoning is distinguished from theoretical reasoning – theoretical reasoning is directed towards beliefs

#### Practical Reasoning

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

#### Intentions in Practical Reasoning

- Intentions pose problems for agents, who need to determine ways of achieving them. If I have an intention to \u03c6, you would expect me to devote
- resources to deciding how to bring about φ.
   Intentions provide a "filter" for adopting other intentions,
- which must not conflict. If I have an intention to  $\phi$ , you would not expect me to adopt
- an intention ψ such that φ and ψ are mutually exclusive.
  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.

#### Intentions in Practical Reasoning

- Agents believe their intentions are possible. That is, they believe there is at least some way that the intentions could be brought about.
- Agents do not believe they will not bring about their intentions.
- It would not be rational of me to adopt an intention to  $\phi$  if I believed  $\phi$  was not possible.
- 6. Under certain circumstances, agents believe they will bring about their intentions. It would not normally be rational of me to believe that I would bring my intentions about; intentions can fail. Moreover, it does not make sense that if I believe φ is inevitable that I would adopt it as an intention.

# Intentions in Practical Reasoning

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

If I believe  $\phi \rightarrow \psi$  and I intend that  $\phi$ , I do not necessarily intend  $\psi$  also. (Intentions are not closed under implication.)

This last problem is known as the *side effect* or *package deal* problem. 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!

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#### Intentions in Practical Reasoning

- Notice that intentions are much stronger than mere desires:
- "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." (Bratman, 1990)

#### Planning Agents

- Since the early 1970s, the AI planning community has been closely concerned with the design of artificial agents
- Planning is essentially automatic programming: the design of a course of action that will achieve some desired goal
- Within the symbolic AI community, it has long been assumed that some form of AI planning system will be a central component of any artificial agent
- Building largely on the early work of Fikes & Nilsson, many planning algorithms have been proposed, and the theory of planning has been well-developed

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#### What is Means-End Reasoning?

- Basic idea is to give an agent:
  - representation of goal/intention to achieve
  - $\hfill\square$  representation actions it can perform
  - representation of the environment
- and have it generate a plan to achieve the goal
- Essentially, this is

automatic programming

















The Blocks	World Operators
<ul> <li>Example 3: The <i>pickup</i> act object <i>x</i> from th</li> </ul>	ion occurs when the arm picks up an ne table.
pre del add	Pickup(x) Clear(x) ~ OnTable(x) ~ ArmEmpty OnTable(x) ~ ArmEmpty Holding(x)
<ul> <li>Example 4: The putdown a object x onto th</li> </ul>	iction occurs when the arm places the ne table.
pre del add	Putdown(x) Holding(x) Holding(x) Clear(x) ^ OnTable(x) ^ ArmEmpty
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The STRIPS approach
The original STRIPS system used a goal stack to control its search
The system has a database and a goal stack, and it focuses attention on solving the top goal (which may involve solving subgoals, which are then pushed onto the stack, etc.)



If on top of goal stack:	Then do:
Compound or single goal matching the current state description	Remove it
Compound goal not matching the current state description	<ol> <li>Keep original compound goal on stack</li> <li>List the unsatisfied component goals or the stack in some <i>new</i> order</li> </ol>
Single-literal goal not matching the current state description	Find rule whose instantiated add-list includes the goal, and 1. Replace the goal with the instantiated rule; 2. Place the rule's instantiated precondition formula on top of stack
Rule	1. Remove rule from stack; 2. Update database using rule; 3. Keep track of rule (for solution)
Nothing	Stop







#### Implementing Practical Reasoning Agents

- So, this agent will have overall optimal behavior in the following circumstances:
- 1. When deliberation and means-ends reasoning take a vanishingly small amount of time; or
- 2. When the world is guaranteed to remain static while the agent is deliberating and performing means-ends reasoning, so that the assumptions upon which the choice of intention to achieve and plan to achieve the intention remain valid until the agent has completed deliberation and means-ends reasoning; or
- When an intention that is optimal when achieved at time  $t_o$  (the time at which the world is observed) is guaranteed to remain optimal until time  $t_c$  (the time at which the agent has found a course of action to achieve the intention).



#### 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



Deliberation	
Agent Control 1	Loop Version 3
1.	-
2. $B := B_0;$	
3. $I := I_0;$	
4. while true	do
5. get nez	xt percept $ ho;$
$6. \qquad B:=brf($	$B, \rho);$
7. $D := opti$	ons(B, I);
8. $I := filter$	(B, D, I);
9. $\pi := plan$	(B,I);
10. $execute(\pi$	•)
11. end while	

![](_page_5_Picture_1.jpeg)

# Commitment Strategies The following commitment strategies are commonly discussed in the literature of rational agents: Biind 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 agent will maintain an intention as long as it is still believed possible.

#### **Commitment Strategies**

- An agent has commitment both to ends (i.e., the 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

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Age 1. 2. 3. 4. 5. 6. 7. 8. 9. 10. 11. 12. 13. 14. 15. 16. 17. 18. 16. 17. 18. 19. 19. 19. 10. 11. 11. 12. 13. 14. 15. 15. 15. 16. 17. 18. 19. 10. 11. 12. 13. 14. 15. 15. 15. 15. 15. 15. 15. 15	nt Control Loop Version 4 $B := B_0;$ $I := I_0;$ while true do get next percept $\rho;$ $B := br(B_i\rho);$ $D := options(B_if);$ $I := filter(B_i, D_i);$ while not empty(\$\pi\$) do $\alpha := hd($\pi$);$ $\pi := tail($\pi$);$ $get next percept $\rho$; B := br(B_i, D_i); if not sound($\pi$, I, B] then\pi := plan(B_i, f)end-ifend-while$	
19. 20.	end-While end-while	
		4-35

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# Commitment Strategies Still overcommitted to intentions: Never stops to consider whether or not its intentions are appropriate Modification: stop to determine whether intentions have succeeded or whether they are impossible: (*Single-minded commitment*)

Agent Control Loop Version 5 2 $B = B_{a}$	
2. D. – D()	
$5. I := I_0;$	
<ol> <li>while true do</li> </ol>	
<ol> <li>get next percept ρ;</li> </ol>	
6. $B := brf(B, \rho);$	
7. $D := options(B, I);$	
8. $I := filter(B, D, I);$	
9. $\pi := plan(B, I);$	
<ol> <li>while not empty(π)</li> </ol>	
or succeeded(I,B)	
or impossible(I,B)) do	
11. $\alpha := hd(\pi)$ ;	
12. $execute(\alpha)$ ;	
13. $\pi := tail(\pi)$ ;	
<ol> <li>get next percept ρ;</li> </ol>	
15. $B := brf(B, \rho)$ ;	
16. if not $sound(\pi, I, B)$ then	
17. $\pi := plan(B, I)$	
18. end-if	
 - 19. end-while	
20. end-while	4-37

![](_page_6_Figure_1.jpeg)

Age	nt Control Loop Version 6	
2	$B := B_{\alpha}$ .	
. 2.		
3.	$I := I_0;$	
4.	while true do	
5.	get next percept $\rho$ ;	
6.	$B := brf(B, \rho);$	
7.	D := options(B, I);	
8.	I := filter(B, D, I);	
9.	$\pi := plan(B, I);$	
10.	while not $(empty(\pi)$	
	or succeeded(I,B)	
	or impossible(I,B)) do	
11.	$\alpha := hd(\pi)$ ;	
12.	execute(a):	
13	$\pi := tail(\pi)$ :	
14	get pert percept a	
14.	get next percept p;	
15.	B := brf(B, p);	
16.	D := options(B, I);	
17.	I := filter(B, D, I);	
18.	if not $sound(\pi, I, B)$ then	
19.	$\pi := plan(B, I)$	
20.	end-if	
21.	end-while	
	end_while	
22.		4-39

![](_page_6_Figure_3.jpeg)

Age:	nt Control Loop Version 7	
2	B := B	
3.	while true do	
4. c	got port porgont a	
5.	gec next percept p,	
· · · ·	$B := \partial f(B, p);$	
1.	D := opnons(B, T);	
8.	T := futer(B, D, T);	
9.	$\pi := plan(B, I);$	
10.	while not $(empty(\pi))$	
	or succeeded(I,B)	
	or impossible(I,B)) do	
11.	$\alpha := hd(\pi)$ ;	
12.	$execute(\alpha)$ ;	
13.	$\pi := tail(\pi)$ ;	
14.	get next percept $\rho$ ;	
15.	$B := brf(B, \rho)$ ;	
16.	if $reconsider(I, B)$ then	
17.	D := options(B, I);	
18.	I := filter(B, D, I);	
19.	end-if	
20.	if not $sound(\pi, I, B)$ then	
21.	$\pi := plan(B, I)$	
22.	end-if	
23.	end-while	
24	end-while	4.41
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![](_page_6_Figure_5.jpeg)

#### Intention Reconsideration

- In situation (1), the agent did not choose to deliberate, and as consequence, did not choose to change intentions. Moreover, if it had chosen to deliberate, it would not have changed intentions. In this situation, the 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 reconsider(...) function is not behaving optimally.
- In situation (3), the agent chose to deliberate, but did not change intentions. In this situation, the *reconsider(...)* function is not behaving optimally.
- In situation (4), the agent chose to deliberate, and did change intentions. In this situation, the reconsider(...) function is behaving optimally.
- An important assumption: cost of *reconsider(...)* is *much* less than the cost of the deliberation process itself.

#### 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,  $\gamma$

#### Optimal Intention Reconsideration

- Results (not surprising):
- If y 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 γ is high (i.e., the environment changes frequently), then cautious agents tend to outperform bold agents. This is because they are able to recognize when intentions are doomed, and also to take advantage of serendipitous situations and new opportunities when they arise.

#### BDI Theory and Practice

- We now consider the semantics of BDI architectures: to what extent does a BDI agent satisfy a theory of agency
- In order to give a semantics to BDI architectures, Rao & Georgeff have developed BDI logics: nonclassical logics with modal connectives for representing beliefs, desires, and intentions
- The 'basic BDI logic' of Rao and Georgeff is a quantified extension of the expressive branching time logic CTL\*
- Underlying semantic structure is a labeled branching time framework

# BDI Logic

- From classical logic: ∧, ₀, ¬, …
- The CTL\* *path quantifiers*:
  - □ A¢ 'on all paths, ¢
- $\square E\phi$  'on some paths,  $\phi$
- The BDI connectives:
- $\square (\mathsf{Bel} \ i \ \phi) \ i \ \mathsf{believes} \ \phi$
- $\Box$  (Des *i*  $\phi$ ) *i* desires  $\phi$
- $\Box$  (Int *i*  $\phi$ ) *i* intends  $\phi$

# BDI Logic

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- Semantics of BDI components are given via accessibility relations over 'worlds', where each world is itself a branching time structure
- Properties required of accessibility relations ensure belief logic KD45, desire logic KD, intention logic KD (Plus interrelationships...)

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![](_page_8_Figure_1.jpeg)

![](_page_8_Figure_2.jpeg)

#### CTL\* Notation

- Variant of branching time logic that we look at is called CTL\*, for Computational Tree Logic (star)
- In this logic
  - □ A = "for every path"
  - E = "there exists a path"
  - □ G = "globally" (similar to □)
  - $\square$  F = "future" (similar to  $\Diamond$ )

#### Paths versus States

- A and E refer to paths
  - A requires that all paths have some property
  - E requires that at least some path has the
- propertyG and F refer to states on a path
- G requires that all states on the given path have some property
- F requires that at least one state on the path has the property

#### CTL\* Examples

- AG p
  - For every computation (i.e., path from the root), in every state, p is true

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- $\hfill\square$  Hence, means the same as  $\Box p$
- EG p

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 There exists a computation (path) for which p is always true

# CTL\* Examples continued

AF p

- □ For every path, eventually state p is true
- $\hfill\square$  Hence, means the same as  $\Diamond p$
- Therefore, p is inevitable

EF p

- There is some path for which p is eventually true
- I.e., p is "reachable"
- □ Therefore, p will hold *potentially*

![](_page_9_Figure_9.jpeg)

#### BDI Logic

- Let us now look at some possible axioms of BDI logic, and see to what extent the BDI architecture could be said to satisfy these axioms
- In what follows, let
   α be an O-formula, i.e., one which contains no
  - positive occurrences of A

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 $\square \phi$  be an arbitrary formula

#### BDI Logic Belief goal compatibility: (Des α) → (Bel α) States that if the agent has a goal to optionally achieve something, this thing must be an option. This axiom is operationalized in the function options: an option should not be produced if it is not believed possible. Goal-intention compatibility: (Int α) → (Des α) States that having an intention to optionally achieve something implies having it as a goal (i.e., there are no intentions that are not goals). Operationalized in the *deliberate* function.

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![](_page_9_Figure_17.jpeg)

#### Implemented BDI Agents: IRMA

- IRMA Intelligent Resource-bounded Machine Architecture - Bratman, Israel, Pollack
- IRMA has four key symbolic data structures: □ a plan library
  - explicit representations of
    - beliefs: information available to the agent may be represented symbolically, but may be simple variables
    - desires: those things the agent would like to make true think of desires as tasks that the agent has been allocated; in humans, not necessarily logically consistent, but our agents will be! (goals)
  - intentions: desires that the agent has chosen and committed to

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### IRMA

- Additionally, the architecture has:
  - □ a *reasoner* for reasoning about the world; an inference engine
  - □ a means-ends analyzer determines which plans might be used to achieve intentions
  - an opportunity analyzer monitors the environment, and as a result of changes, generates new options
  - a filtering process determines which options are compatible with current intentions

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□ a deliberation process responsible for deciding upon the 'best' intentions to adopt

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# SHRDLU Dialog

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