

# Leveraging New Plans in AgentSpeak(PL)

Felipe Meneguzzi

`felipe.meneguzzi@kcl.ac.uk`

Michael Luck

`michael.luck@kcl.ac.uk`

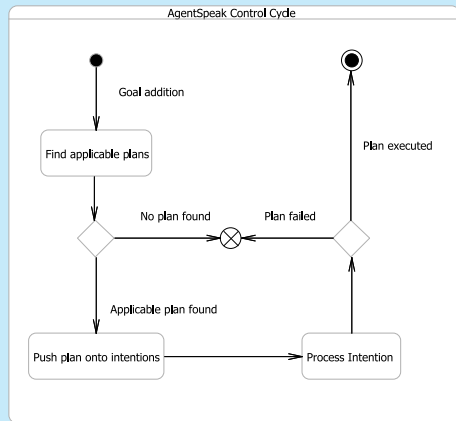
<sup>1</sup>Department of Computer Science  
King's College London

Declarative Agent Languages and Technologies  
AAMAS 2008 – Estoril, Portugal

- 1 Background
- 2 AgentSpeak(PL)
- 3 Leveraging new plans
- 4 Experiments and Results
- 5 Conclusions and Future Work

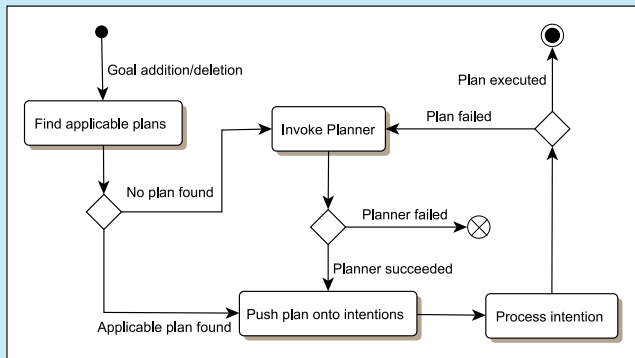
# AgentSpeak(L)

- Procedural agent language
- Based on the BDI model
- Designer specifies plans in a library
  - ▶ Plans encode procedures
  - ▶ Plans are characterised by trigger and context condition
  - ▶ Goals are implicit in the plans



# Planning in AgentSpeak(PL)

- AgentSpeak(L) + Planning
  - ▶ Standard AgentSpeak(L) language
  - ▶ Planner invoked through an atomic action
- In principle, any state-space planner can be used



# AgentSpeak(L) to STRIPS

+!move(O, A, B)	→	operator: move(O, A, B)
: at(O, A) & not at(O, B)	→	pre: at(A) & not at(B)
<- -at(A);	→	del: at(A)
+at(B).	→	add: at(B)

- Relies on clear similarities between AgentSpeak plans and STRIPS operators
- Desired world state becomes the planners goal
- Belief base becomes the planners start state

# STRIPS to AgentSpeak(L)

STRIPS plan to achieve

```
battery(full):
```

```
move(1,1)
```

```
move(1,2)
```

```
charge
```

AgentSpeak(L):

```
+!goal_conj([battery(full)]
```

```
: true
```

```
<- !move(1,1);
```

```
!move(1,2);
```

```
!charge.
```

- Each STRIPS action correspond to a low-level AgentSpeak(L) plan
- Plans amount to a series of AgentSpeak(L) subgoals

# Executing Plans and Limitations

- Generated plan is executed as a regular AgentSpeak(L) plan
- Planning is computationally expensive
- New plans should be added to the plan library
- However this is not so trivial:
  - ▶ How should a new plan be added?
  - ▶ What should the context condition be?

# Leveraging new plans

- Key aspects:
  - ▶ Ordering of the plan library - new plans must come before 'planning plan'
  - ▶ Generation of minimum context condition



# Ordering Example - Pseudo PL

```
+!move(P,A,B) : empty(B) & over(P,A)  
  <- ...
```

```
+!process(P,A) : over(P,A)  
  <- ...
```

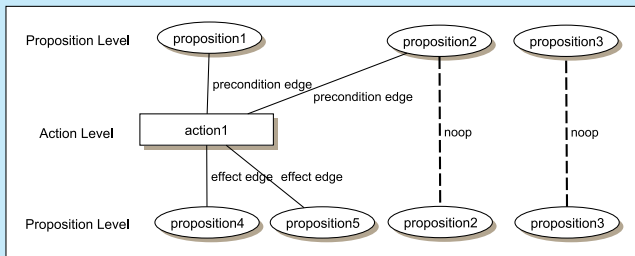
```
+!goal_conj([over(p1,u1)]) : over(p1,u2)  
  <- !move(p1,u2,u1).
```

```
// Place new plans here
```

```
+!goal_conj(Goals) : true  
  <- .plan(Goals).
```

# Planning Graph

- Context generation algorithm uses planning graph
- Directed levelled graph
- Interleaved proposition and action levels
- Preconditions and Effect arcs connect levels



# Generating Context Information

- Preconditions of a plan step must be true earlier in the graph
- Need to propagate preconditions back to previous operators, or to the first level
- Intuitively:
  - ▶ Create a planning graph with the target plan
  - ▶ Connect preconditions of each action level to the previous one
  - ▶ If no action causes the precondition in that level, add a *noop*
  - ▶ Propositions propagated to the first level become the context condition

# A Production Cell Example

- Production cell with four processing units and a crane
- Parts can be moved around to processing units
- Processing units can process parts

Operator	Preconditions	Effects
<code>move (P, A, B)</code>	<code>empty (B)</code> <code>over (P, A)</code>	$\sim$ <code>empty (B)</code> $\sim$ <code>over (P, A)</code> <code>over (P, B)</code> <code>empty (A)</code>
<code>process (P, A)</code>	<code>over (P, A)</code>	<code>processed (P)</code>

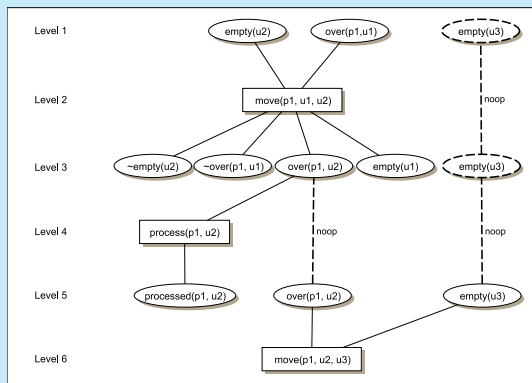
# A Production Cell Example: Part II

Generating context  
condition for plan:

`move(p1, u1, u2)`

`process(p1, u2)`

`move(p1, u2, u3)`

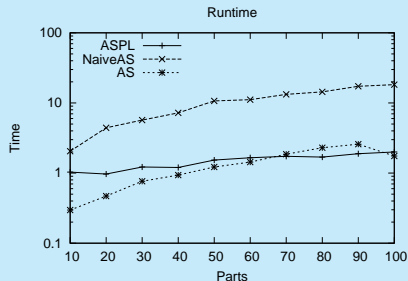


# Experiments

- Uses production cell scenario
- Large numbers of parts coming in for processing
- Three types of parts, different sequences of processing
- Measure agent reaction time as new plans are created
- AgentSpeak(L) versus Naive AgentSpeak(PL) versus AgentSpeak(PL)

# Results

- Naive AgentSpeak(PL) very inefficient
- Plan reuse strategy amortises cost of planning
- Over time, computational cost of planning approximates traditional AgentSpeak(L)



# Conclusions and Future Work

- Conclusions:
  - ▶ Plan reuse bridges performance gap introduced by planning
  - ▶ Algorithm has polynomial complexity (like graph construction in GraphPlan)
  - ▶ More complex than necessary, but extensible
- Future Work:
  - ▶ Extending algorithm to handle richer operators



# Questions?

# Surprise Slide - Alternative algorithm

- A simpler algorithm may be possible

$Open = \emptyset$

**for**  $i = n$  to 2 **do**

$Open = Open \cup preconditions(a_i)$

$Open = Open - postcondition(a_i - 1)$

**end for**  $Open = Open \cup preconditions(a_1)$