## Probabilistic BDI Agents: Actions, Plans, and Intentions

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Definition:

An entity

which **perceives** its situated environment, which **deliberates** accordingly, which **takes actions** autonomously,

in order to achieve its design objectives



Michael Wooldridge. An Introduction to Multiagent Systems. John Wiley and Sons Ltd, February 2002

#### Conceptual Framework

An entity

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#### Belief-Desire-Intention (BDI) Framework



#### Belief-Desire-Intention Framework

Overview



#### Belief-Desire-Intention (BDI) Framework



Question











Summary



Summary



our approach

Summary



Summary



our approach

Limitation



Limitation



Our approach



Our approach

Quantitative properties analysis

probabilistic effects of an action

plans

actions

which plan to select if there is more than one available



which intention to progress if there is more than one available

#### Building Trustworthy BDI Agents Actions

original BDI semantics<sup>1</sup>

$$\frac{act: \varphi \leftarrow \langle \phi^-, \phi^+ \rangle \quad \mathcal{B} \vDash \varphi}{\langle \mathcal{B}, act \rangle \rightarrow \langle (\mathcal{B} \setminus \phi^- \cup \phi^+), nil \rangle}$$

act

probabilistic BDI semantics

$$\frac{act: \varphi \leftarrow \mu \quad \mu(\phi^-, \phi^+) = p \quad \mathcal{B} \vDash \varphi}{\langle \mathcal{B}, act \rangle \rightarrow_p \langle (\mathcal{B} \backslash \phi^- \cup \phi^+), nil \rangle} \qquad act^p$$

 $\mu = [(\phi_1^-, \phi_1^+) \mapsto p_1, \cdots, (\phi_n^-, \phi_n^+) \mapsto p_n] \text{ and } \sum_{i=1}^n p_i = 1.$ 

<sup>1</sup>CAN Semantics in [Winikoff et al., 2002]

Plans

original BDI semantics

$$\varphi: P \in \Delta \quad \mathcal{B} \vDash \varphi$$
  
$$\langle \mathcal{B}, e: (|\Delta|) \rangle \rightarrow \langle \mathcal{B}, P \succ e: (|\Delta \setminus \{\varphi: P\}|) \rangle$$

select

probabilistic BDI semantics

$$\frac{\varphi: P \in \Delta \quad \delta^{p}(\mathcal{B}, \Delta) = \mu \quad \mu \neq \perp \quad \mu(\varphi: P) = p}{\langle \mathcal{B}, e: (|\Delta|) \rangle \rightarrow_{p} \langle \mathcal{B}, P \rhd e: (|\Delta \setminus \{\varphi: P\}|) \rangle}$$

 $select^p$ 

probabilistic plan selection function:  $\delta^p: 2^{\mathcal{B}} \times 2^{\Pi} \rightarrow Dist(\Pi) \cup \{\bot\}$ 

Intentions

original BDI semantics  

$$\frac{P \in \Gamma \quad \langle \mathcal{B}, P \rangle \rightarrow \langle \mathcal{B}', P' \rangle}{\langle E^{e}, \mathcal{B}, \Gamma \rangle \Rightarrow \langle E^{e}, \mathcal{B}', (\Gamma \setminus \{P\}) \cup \{P'\} \rangle} \qquad A_{step} \\
\frac{P \in \Gamma \quad \langle \mathcal{B}, P \rangle \rightarrow}{\langle E^{e}, \mathcal{B}, \Gamma \rangle \Rightarrow \langle E^{e}, \mathcal{B}', \Gamma \setminus \{P\} \rangle} \qquad A_{update} \\
\frac{P \in \Gamma \quad \eta^{p}(\mathcal{B}, \Gamma) = \mu \quad \mu \neq \perp \quad \mu(P) = p \quad \langle \mathcal{B}, P \rangle \rightarrow_{p'} \langle \mathcal{B}', P' \rangle}{\langle E^{e}, \mathcal{B}, \Gamma \rangle \Rightarrow_{p \cdot p'} \langle E^{e}, \mathcal{B}', (\Gamma \setminus \{P\}) \cup \{P'\} \rangle} \qquad A_{step}^{p} \\
\frac{P \in \Gamma \quad \eta^{p}(\mathcal{B}, \Gamma) = \mu \quad \mu \neq \perp \quad \mu(P) = p'' \quad \langle \mathcal{B}, P \rangle \rightarrow_{1}}{\langle E^{e}, \mathcal{B}, \Gamma \rangle \Rightarrow_{p''} \langle E^{e}, \mathcal{B}, \Gamma \setminus \{P\} \rangle} \qquad A_{update}^{p} \\
\frac{P \in \Gamma \quad \eta^{p}(\mathcal{B}, \Gamma) = \mu \quad \mu \neq \perp \quad \mu(P) = p'' \quad \langle \mathcal{B}, P \rangle \rightarrow_{1}}{\langle E^{e}, \mathcal{B}, \Gamma \rangle \Rightarrow_{p''} \langle E^{e}, \mathcal{B}, \Gamma \setminus \{P\} \rangle} \qquad A_{update}^{p} \\
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\frac{P \in \Gamma \quad \eta^{p}(\mathcal{B}, \Gamma) = \mu \quad \mu \neq \perp \quad \mu(P) = p'' \quad \langle \mathcal{B}, P \rangle \rightarrow_{1}}{\langle \mathcal{B}, P \rangle} \qquad A_{update}^{p} \\
\frac{P \in \Gamma \quad \eta^{p}(\mathcal{B}, \Gamma) = \mu \quad \mu \neq \perp \quad \mu(P) = \mu \quad \langle \mathcal{B}, P \rangle \rightarrow_{1}}{\langle \mathcal{B}, P \rangle} \qquad A_{update}^{p} \\
\frac{P \in \Gamma \quad \eta^{p}(\mathcal{B}, \Gamma) = \mu \quad \mu \neq \perp \quad \mu(P) = \mu \quad \langle \mathcal{B}, P \rangle}{\langle \mathcal{B}, P \rangle} \rightarrow_{1}} \qquad A_{update}^{p} \\
\frac{P \in \Gamma \quad \eta^{p}(\mathcal{B}, \Gamma) = \mu \quad \mu \neq \perp \quad \mu(P) = \mu \quad \mu \neq \perp \quad \mu(P) = \mu \quad \mu \neq \perp \quad \mu(P) \in \mu \quad \mu \neq \perp \quad \mu(P) \in \mu \quad \mu \neq \mu \quad \mu(P) \in \mu \quad \mu \quad \mu(P) \in \mu \quad \mu(P) \in \mu \quad \mu \quad \mu(P) \in \mu \quad$$

probabilistic intention selection function:  $\delta^p: 2^{\mathcal{B}} \times 2^{\Gamma} \to Dist(\Gamma) \cup \{\bot\} \text{ and } \sum_{p''} p'' + \sum_{p,p'} p \cdot p' = 1$ 

Summary on distribution

Action:

 $\mu = [(\phi_1^-, \phi_1^+) \mapsto p_1, \cdots, (\phi_n^-, \phi_n^+) \mapsto p_n] \text{ and } \sum_{i=1}^n p_i = 1.$ 

pre-designed distributions

Summary on distribution

Action: $\mu = [(\phi_1^-, \phi_1^+) \mapsto p_1, \cdots, (\phi_n^-, \phi_n^+) \mapsto p_n]$  and  $\sum_{i=1}^n p_i = 1$ .pre-designed distributionsPlan: $\delta^p: 2^{\mathcal{B}} \times 2^{\Pi} \rightarrow Dist(\Pi) \cup \{\bot\}$ situational distributions

Intention:  $\delta^p: 2^{\mathcal{B}} \times 2^{\Gamma} \to Dist(\Gamma) \cup \{\bot\}$ 

situational distributions

Summary on distribution

Action:  $\mu = [(\phi_1^-, \phi_1^+) \mapsto p_1, \cdots, (\phi_n^-, \phi_n^+) \mapsto p_n] \text{ and } \sum_{i=1}^n p_i = 1.$ pre-designed distributions Plan:  $\delta^p: 2^B \times 2^\Pi \to Dist(\Pi) \cup \{\bot\}$ situation value description situational distributions  $\langle d_0, \{(\varphi_1, d_1), \cdots, (\varphi_n, d_n)\}, f \rangle$ Intention:  $\delta^p: 2^B \times 2^\Gamma \to Dist(\Gamma) \cup \{\bot\}$ default value  $d_0$ , if  $\varphi_i$  holds, it will be aggregated according to the function f

situational distributions

Evaluation



Evaluation



Evaluation



Evaluation



probabilistic effects of action

#### Analysis

#### Table 2. Plan and intention selection strategies. production line storage line Plan Selection Strategies Intention Selection Strategies **SMP**: Select Most Preferred SMU: Select Most Urgent PSD: Preference Situational Distribution FIFO: First-In-First-Out **RR**: Round Robin **PUSD**: Pure Urgency Situational Distri-bution LUSD: Layered Urgency Situational Distribution **OLUSD**: Optimised Layered Urgency Sit-uational Distribution

#### Analysis

#### production line

Table 2. Plan and intention selection strategies.

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Plan Selection Strategies	Intention Selection Strategies		
SMP: Select Most Preferred	SMU: Select Most Urgent		
<b>PSD</b> : Preference Situational Distribution	FIFO: First-In-First-Out		
	RR: Round Robin		
	<b>PUSD</b> : Pure Urgency Situational Distri-		
	bution		
	LUSD: Layered Urgency Situational Dis-		
	tribution		
	<b>OLUSD</b> : Optimised Layered Urgency Sit-		
	uational Distribution		

**Table 3.** Probability of product 1, product 2 for the properties, *e.g.* (S1, S2) with different plan and intention selection strategies listed in Table 2.

	SMU		FIFO		RR		
P l a n	S M P	(S1, S2)	(S1,F2)	(S1, S2)	(S1,F2)	(S1, S2)	(S1, F2)
		0	0.9	0	0	0	0
		(F1,S2)	(F1, F2)	(F1,S2)	(F1, F2)	(F1, S2)	(F1, F2)
		0	0.1	0.9	0.1	0	1
	P S D	(S1, S2)	(S1,F2)	(S1, S2)	(S1,F2)	(S1, S2)	(S1,F2)
		0	0.93	0	0	0	0
		(F1,S2)	(F1, F2)	(F1,S2)	(F1, F2)	(F1,S2)	(F1, F2)
		0	0.07	0.93	0.07	0	1
PUSD			LUSD		OLUSD		
P l a n	G	(S1,S2)	(S1,F2)	(S1, S2)	(S1,F2)	(S1, S2)	(S1,F2)
	M	0.03	0.48	0.510	0	0.97	0
	D	(F1,S2)	(F1, F2)	(F1,S2)	(F1, F2)	(F1,S2)	(F1, F2)
	1	0.08	0.41	0.482	0.008	0.037	0
	D	(S1, S2)	(S1,F2)	(S1, S2)	(S1,F2)	(S1, S2)	(S1,F2)
	G	0.03	0.49	0.513	0	0.98	0
	D	(F1,S2)	(F1, F2)	(F1, S2)	(F1, F2)	(F1, S2)	(F1, F2)
	D	0.08	0.4	0.481	0.05	0.02	0

#### storage line

#### Analysis

#### Table 2. Plan and intention selection strategies. production line Plan Selection Strategies Intention Selection Strategies **SMP**: Select Most Preferred SMU: Select Most Urgent **PSD**: Preference Situational Distribution **FIFO**: First-In-First-Out **RR**: Round Robin **PUSD**: Pure Urgency Situational Distri-bution LUSD: Layered Urgency Situational Distribution **OLUSD**: Optimised Layered Urgency Sit-uational Distribution Table 4. DTMC generation: final size and timing. Strategies States Transitions Build time (s) Rule Applications (SMP, SMU) 31 30 66.57 217 31 (SMP, FIFO) 30 65.85 211 (SMP, RR)19 18 52.13143(PSD, SMU) 36 36 92.2627336 36 92.25(PSD, FIFO) 26819 18 51.72(PSD, RR)143572845 2447.37 5300(SMP, PUSD) 323 (SMP, LUSD) 478 1518.363116 323 1481.07 3116 (SMP, OLUSD) 478 (PSD, PUSD) 697 1039 17435.90 6836 (PSD, LUSD) 417 614 2106.644157(PSD, OLUSD) 417 614 2098.514157

#### storage line

#### Analysis



Fig. 2. Probability of reaching the end state (product 1, product 2) with increasing failure probability in (PSD, SMU) and (PSD, OLUSD).

Note



In principle, our approach allows users to "run" any example with

- 1. different external events (e.g. tasks)
- 2. different plan libraries (e.g. methods)
- 3. custom situation value descriptions

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provide probabilistic support

to deliver quantitative assurance of agent behaviours

in more realistic industrial systems prone to failures

#### Building Trustworthy BDI Agents Future Work

- 1. multi-objective analysis: obtaining high success rate while keeping the overall bag cost low
- 2. dynamic environment: new product being reduced and more robots being employed



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## Many thanks for your attentions



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