

Social Performance Measurement for Multiagent Systems

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Motivation

- Multiagent systems (MAS) find their way to industrial and commercial applications
- Measuring their performance becomes increasingly important
- Agent-oriented Software Engineering (AOSE) provides no methods yet
- Motivation for *social performance measurement*:
 - challenge: complexity (decentralisation, emergence, micro-macro)
 - opportunity: exploiting high-level communication between agents
- This talk: first steps and initial results

Overview

Introduction

Introduction

Software Engineering is (among other things) concerned with meeting product requirements, such as

- availability,
- modifiability,
- security,
- usability,
- . . .

In *Performance Engineering*, ideally,

- *external attributes* should be derived from
- *internal attributes*

Introduction (II)

In MAS particularly challenging due to

- encapsulation of functionality at agent-level,
 - adaptive agents in dynamic environments,
 - complex interactions between computational sub-processes,
 - emergent (unforeseen) macro-level phenomena,
 - potential ignorance of agents' internal functionalities.
- lack of work on performance measurement

Introduction (III)

Observation:

MAS exhibit certain properties that are actually *advantageous* in the development of measures of internal attributes.

1. They are (usually) based on deliberative, knowledge-based (i.e. symbolically operating) agents.
2. Agents in MAS (usually) communicate in high-level languages such as KQML or FIPA-ACL.

Principles of Social Performance Measurement

Social Performance Measurement (I)

Idea of social performance measurement: To exploit

- (a) the existing *layer of ongoing, symbolic communication* in the system which
- (b) captures the *interactions* between agents
- (c) and which is *comprehensible* for humans.

Goals:

- develop simple, easily measurable measures based solely on internal communication data (\Rightarrow social)
- map these to external attributes

Social Performance Measurement (II)

Assumptions:

- MAS exhibit *social properties* that come about from communication processes
- use *communication data* as data material for performance measurement

Communication can be

1. a (textual) message passed between two or more agents or
2. some (physical) action that an agent performs publicly.

Social Performance Measurement (III)

Advantages:

1. By measuring communication we abstract from non-communicative properties of the system
→ immense reduction of global system complexity
2. We can exploit knowledge-level representations between knowledge-based agents
3. We are able to measure performance in *open systems* (where internal agent data cannot be accessed)

Prerequisite for defining measures: definition of an underlying MAS model.

A Generic MAS Model

A Generic MAS Model

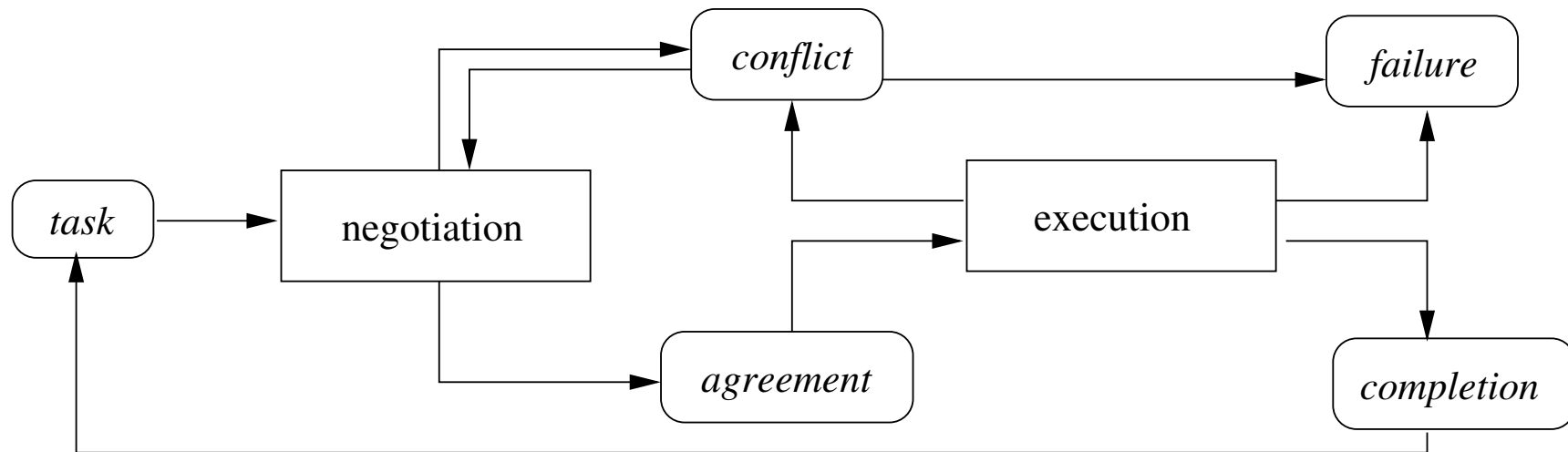
- MAS consists of a set of agents $\mathcal{A} = \{a, b, c, \dots\}$ capable of communicating in a high-level ACL (here: FIPA-ACL)
- Semantics of the ACL are common knowledge and obeyed (benevolence is *not* implied by this)
- MAS has to perform tasks from \mathcal{T} that enter the system at arbitrary points in time
- A real-valued measure

$$d : \mathcal{T} \rightarrow \mathbf{R}$$

for the difficulty of the tasks is available

A Generic MAS Model

Top-level processing cycle:



Communication (I)

Subset of FIPA-ACL messages of the form

$$\text{performative}(\text{sender}, \text{receiver}, \text{content})$$

where

$$\text{performative} \in \{ \text{inform}, \text{inform_done}, \text{inform_re}, \\ \text{agree}, \text{accept_proposal}, \text{request}, \\ \text{cfp}, \text{reject_proposal}, \text{propose}, \\ \text{failure}, \text{not_understood}, \text{refuse} \}$$

We measure all quantities wrt *total number of messages*

$TNOM = |M|$ where the set of all messages is

$$M = \{m_1, m_2, \dots, m_n\}$$

Communication (II)

Further definitions:

Partitioning messages according to tasks in

$$M_T = \{m_{T_1}, \dots, m_{T_{n_i}}\}$$

where $T \in \mathcal{T}_{curr}$ (set of processed tasks = $\mathcal{T}_{curr} \subseteq \mathcal{T}$)
such that

$$M = \biguplus_{T \in \mathcal{T}_{curr}} M_T.$$

Partition tasks into “successfully completed” and
“failed” tasks:

$$\mathcal{T}_{curr} = \mathcal{T}_{succ} \uplus \mathcal{T}_{failed}$$

Defining Social Performance Measures

Basic Performance Measures (I)

Very simple: counting messages and message types

1. *MPTD=messages per task and difficulty:*

$$MPTD = \sum_{T \in \mathcal{T}_{curr}} \frac{M_T}{d(T)}$$

2. *fail-fast* version ($0 \leq \alpha \ll \beta \leq 1$):

$$ff_MPTD = \alpha \cdot \sum_{T \in \mathcal{T}_{succ}} \frac{M_T}{d(T)} + \beta \cdot \sum_{T \in \mathcal{T}_{failed}} \frac{M_T}{d(T)}$$

Basic Performance Measures (II)

Less trivial: we can define *types* of performatives.
Assume partition

- $Type1 = \{\text{request}\},$
- $Type2 = \{\text{inform, inform_done, inform_ref}\}$
- $Type3 = \{\text{cfp, propose}\}$
- $Type4 = \{\text{reject_proposal, refuse}\}$
- $Type5 = \{\text{accept_proposal, agree}\}$
- $Type6 = \{\text{not_understood, failure}\}$

→ We can count occurrences of these types

Basic Performance Measures (III)

We obtain as further measure the *mean message type usage MMTU*:

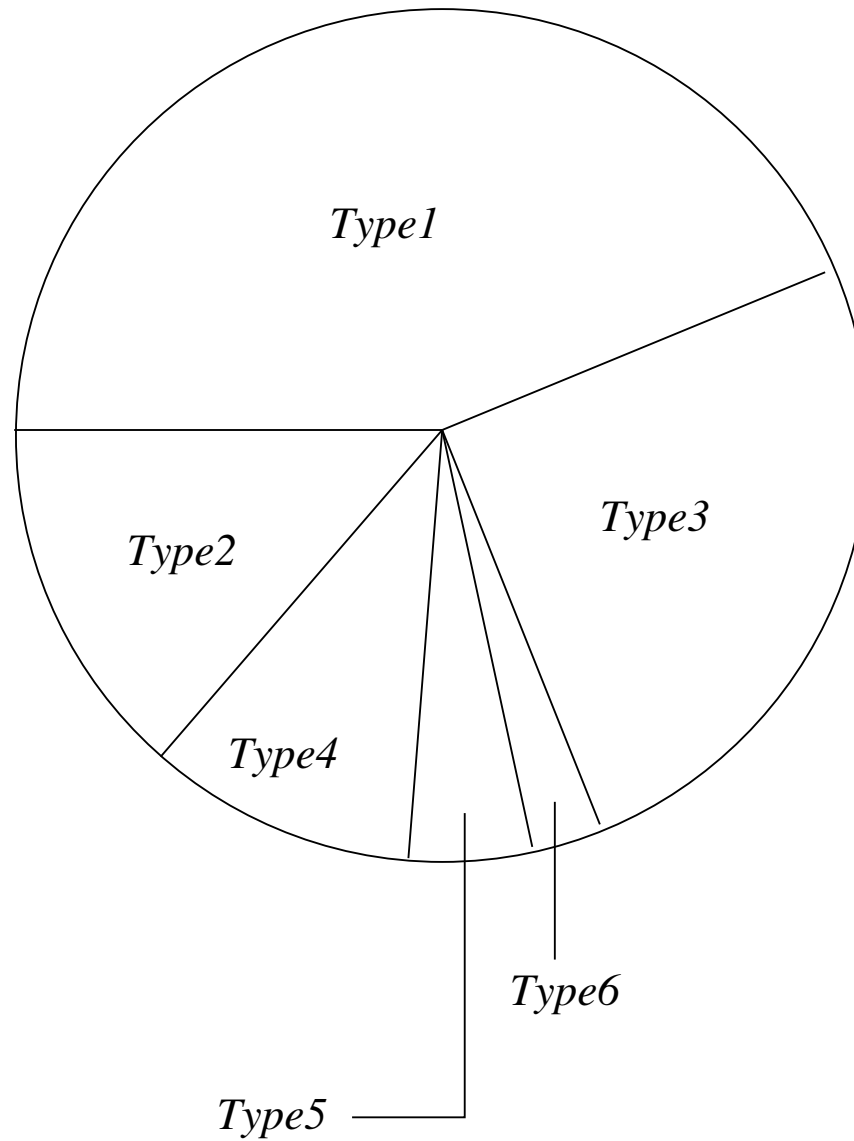
$$MMTU(x) = \frac{1}{|\mathcal{T}_{curr}|} \sum_{T \in \mathcal{T}_{curr}} \frac{|\{m \in M_i | type(m) = x\}|}{|M_T|}$$

for $x \in \{Type1, \dots, Type6\}$ Trivially *MMTU* subsumes *mean performative usage*.

Visualisation: *message type partition chart*

Basic Performance Measures (IV)

Example:



Basic Performance Measures (IV)

MMTU can be refined through parametrisation, e.g.

$$MMTU(x, d_1, d_2) = \frac{1}{|\mathcal{T}_{curr}[d_1 : d_2]|} \sum_{T \in \mathcal{T}_{curr}[d_1 : d_2]} \frac{|\{m \in M_i \mid type(m) = x\}|}{|M_T|}$$

where

$$\mathcal{T}_{curr}[d_1 : d_2] = \{T \in \mathcal{T}_{curr} \mid d_1 \leq d(T) \leq d_2\}$$

Alternatives to difficulty: subsets of agents/agent types, spatial network regions, etc.

Complex Message Patterns (I)

- Simple measures don't allow for *syntactic analysis* of communication
- We introduce *message patterns* to abstract from message sequences
- Use variables $A, B, C \dots$ for messages or message fields (participant, content) and \star as a wildcard symbol
- *Examples:*
$$p = [\text{accept}(A, B, \text{do}(A, X)), *, \text{do}(A, X)]$$
$$q = [\text{accept}(A, B, \text{do}(A, X)), (\neg \text{do}(A, X))^n]$$
- Define $\text{matches}(m, p)$ as a boolean function that matches message sequence m against pattern p

Complex Message Patterns (II)

Pattern-based measures:

- Average length of p :

$$mean_length(p) = \frac{1}{|\mathcal{T}_{curr}|} \sum_{matches(m,p) \wedge m \in M} length$$

- Task-relative version:

$$mean_length(p) = \frac{1}{|\mathcal{T}_{curr}|} \sum_{matches(m,p) \wedge m \in M} \frac{length}{|M|}$$

- Frequency of occurrence:

$$frequency(p) = \frac{|\{m \in M \mid matches(m, p)\}|}{|M|}$$

Complex Message Patterns (III)

Examples:

Let

$$p_a = \bigcap_{Q \in \mathcal{A}} [\text{propose}(P, Q, X), *, \text{accept}(Q, P, X)],$$

$$p_c = \bigcup_{Q \in \mathcal{A}} [\text{propose}(P, Q, X), *, \text{reject_proposal}(Q, P, X)]$$

Then *mean time to agreement (MTTA)* and *mean time to conflict (MTTC)* can be defined as

$$MTTA = \text{mean_length}(p_a),$$

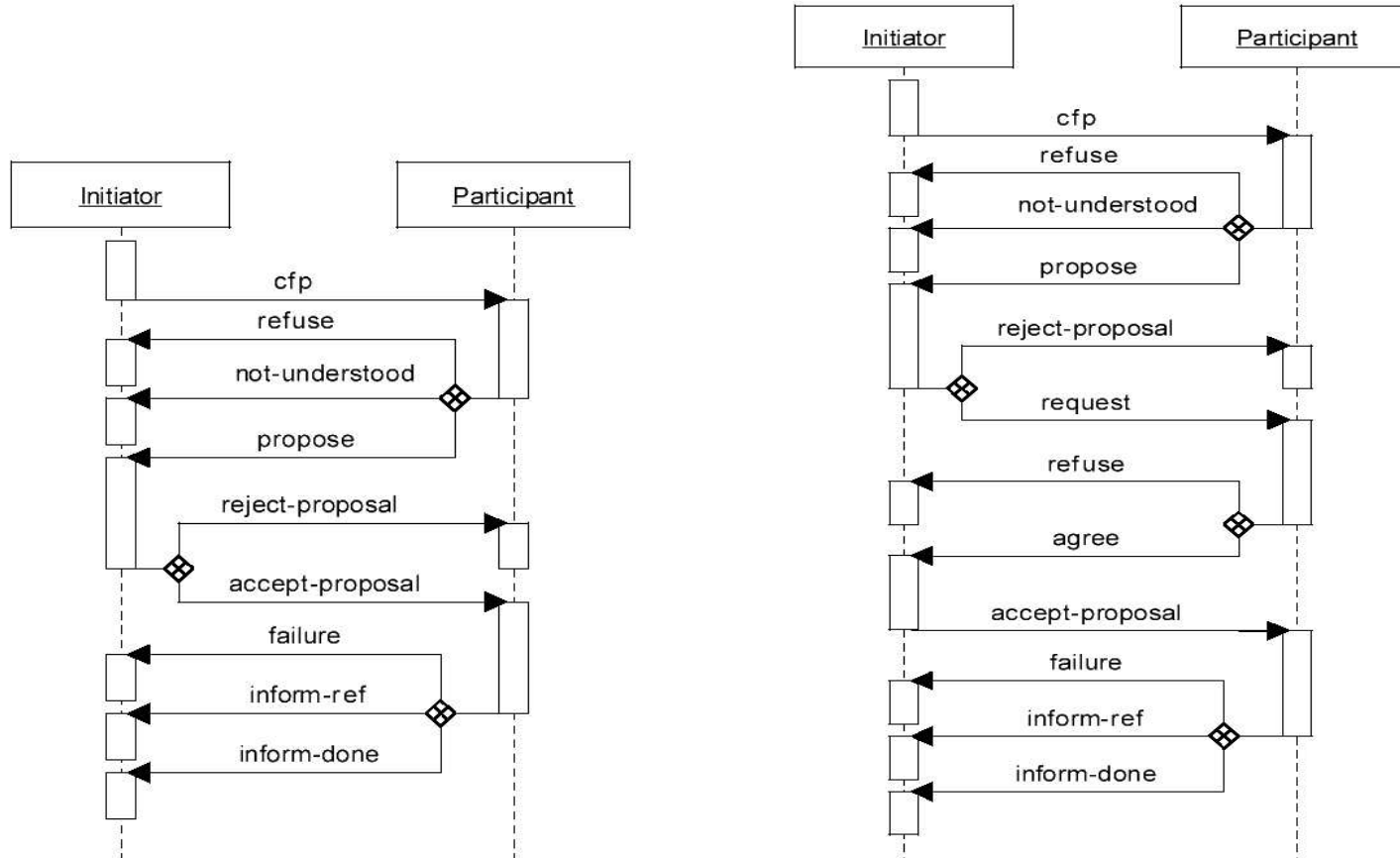
$$MTTC = \text{mean_length}(p_c)$$

→ Usefulness of measure definitions in applications depends on *interaction protocols*

Example

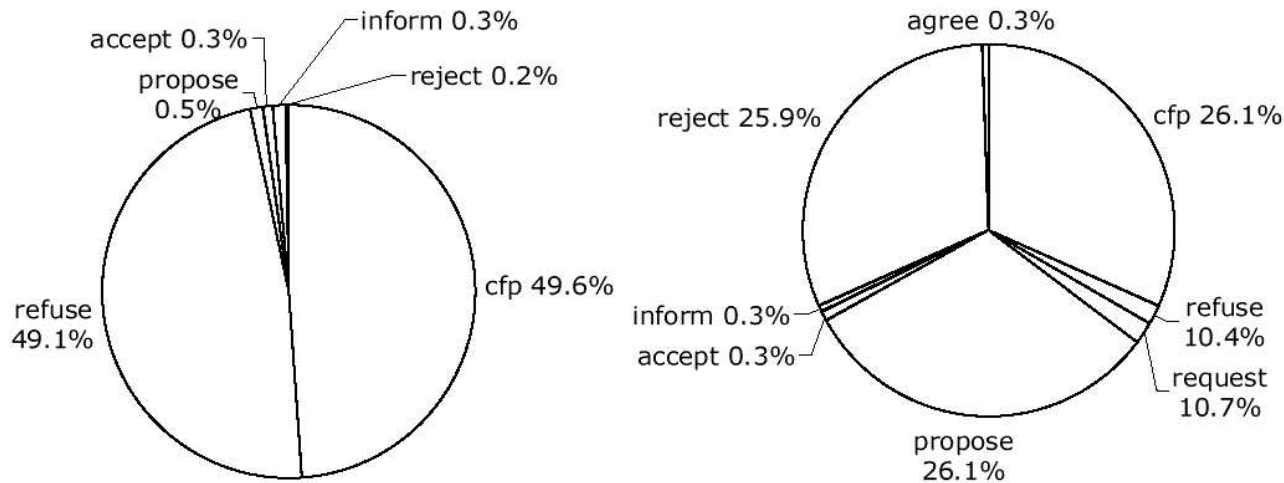
Example (I)

Comparison of the Contract-Net Protocol with the Contract-Net-With-Confirmation Protocol:



Example (II)

- Problem: find internal attribute that best predicts *assigned tasks ratio* (ATR) where $ATR(CNP)=0.65$ and $ATR(CNCP)=1.0$
- Evaluation of *TNOM* (20130/30768 resp.) offers no explanation
- Neither does evaluation of *MMTU*:



Example (III)

- Therefore, we measure the frequency of “rejected unconditional proposals”

$p = [\text{propose}(A, B, X, \{\}), *, \text{reject_proposal}(B, A, X)]$

- Result: $\text{frequency}(p, \text{CNP}) = 0.35$,
 $\text{frequency}(p, \text{CNCP}) = 0.0$
- Superiority of CNCP is due to avoidance of “hasty allocations”
- Complex measure explained what simple measures couldn’t explain

Conclusions & Outlook

Summary

- We suggested *social performance measures* for MAS based on measuring communication processes
- Claimed that these aid complexity reduction while correctly predicting *external* from *internal attributes*
- Main feature: abstraction from intra-agent reasoning processes and physical execution
- Exploiting knowledge-level representations intelligent agents use
- Defined concrete measures, illustrated usefulness with examples

Outlook (I)

- Development and critical evaluation of further generic measures (in particular, standardised patterns)
- Further refinement of measures (advanced statistical methods)
- Definition of qualitative *semantic measures*:
 - classification into processing phases
 - analysing social relationships
 - logical properties - view communication as
 - distributed search
 - constraint satisfaction processes
 - distributed theorem proving

Outlook (II)

- Investigation of relationship external \leftrightarrow internal attributes
- Explore combination of social-level and cognitive-level measurements
- Explore more complex social phenomena (group formation, socialisation, power, authority, etc.)
- Develop tools for automatic measurement of complex MAS

Thank you for your attention!