

# Two Forms of Explanations in Computational Assumption-based Argumentation

## (Extended Abstract)

Xiuyi Fan, Siyuan Liu, Huiguo Zhang,  
Chunyan Miao  
Joint NTU-UBC Research Centre of Excellence  
in Active Living for the Elderly (LILY)  
Nanyang Technological University, Singapore

Cyril Leung  
Electrical and Computer Engineering  
The University of British Columbia, Canada

### ABSTRACT

Computational Assumption-based Argumentation (CABA) has been introduced to model argumentation with numerical data processing. To realize the “explanation power” of CABA, we study two forms of argumentative explanations, *argument explanations* and *CU explanations* representing *diagnosis* and *repair*, resp.

### Keywords

Argumentation, Explanation

## 1. INTRODUCTION

Assumption-based Argumentation (ABA) [8] is a form of structured argumentation with applications in many areas [6]. However, when used as a modeling tool, ABA has limited ability to directly model systems involving numerical calculation. For instance, in ABA based decision making work, e.g. [3, 4], the relations between decision candidates and agent goals need to be “pre-compiled” into binary predicates rather than analyzed from data. The lack of numerical calculation is a major hindrance to ABA applications requiring intensive data processing.

The Computational Assumption-based Argumentation (CABA) framework [2], an ABA extension, introduced Computation Units (CUs) [5] to capture computation that is difficult to represent with standard ABA. A unique advantage of CABA is that, while supporting numerical calculation, it enhances the “explanation power” of argumentation by connecting results obtained from numerical calculation to high-level arguments. We study two forms of CABA explanations, *argument explanation (arg-explanation)* and *CU-explanation*, for non-acceptable arguments. We leverage on the established relation between CABA and Abstract Argumentation (AA) [1] for our work. For a non-acceptable argument  $A$ , its arg-explanation gives a form of *diagnosis*, identifying attacking arguments that cannot be defended. Its CU-explanation represents a form of *repair*, identifying “fixes” that would render  $A$  acceptable.

## 2. EXPLANATION IN CABA

We introduce CABA explanations with a version of the Multiple Attribute Decision Making problem presented in [9]. *Good Col-*

**Appears in:** *Proc. of the 16th International Conference on Autonomous Agents and Multiagent Systems (AAMAS 2017)*, S. Das, E. Durfee, K. Larson, M. Winikoff (eds.), May 8–12, 2017, São Paulo, Brazil.  
Copyright © 2017, International Foundation for Autonomous Agents and Multiagent Systems (www.ifaamas.org). All rights reserved.

**Table 1: Student Candidate Admission Data.**

Student	Exam <sub>1</sub>	Exam <sub>2</sub>	Interview	EA
s1	92	89	A	No
s2	93	85	A	No

lege is admitting students. To evaluate candidates, four attributes are considered: *Exam<sub>1</sub>*, *Exam<sub>2</sub>*, *Interview* and *Extracurricular Activity (EA)*. *Exam<sub>1</sub>* and *Exam<sub>2</sub>* are scores ranging from 0 to 100; *Interview* is a rank from *E* to *A*; *EA* is a binary value, (*Yes/No*). The selection criterion is specified with two conditions C1 and C2, such that: (C1) The average score of *Exam<sub>1</sub>* and *Exam<sub>2</sub>* is greater than 90, or *EA* is *Yes*; and (C2) the *Interview* rank is *A*. A student is admitted iff both C1 and C2 are met.

Table 1 presents the attributes of two candidates,  $s_1$  and  $s_2$ . Here, we can see that for student  $s_1$ , his average exam score is  $(92 + 89)/2 = 90.5$ , hence meeting condition C1; his interview rank is *A*, meeting condition C2; therefore  $s_1$  should be admitted. For  $s_2$ , his average exam score is  $(93 + 85)/2 = 89$  and he has not performed any extracurricular activity, thus failing to meet C1; although  $s_2$  has an *A* for his interview,  $s_2$  cannot be admitted. Here, we need to compute the average scores of *Exam<sub>1</sub>* and *Exam<sub>2</sub>* and test if the average is greater than 90. We pack this computation into a CU,  $u_{90} = \langle T_{90}, C_{90}, E_{90} \rangle$ , in which:

- $T_{90} \subseteq \mathbb{Z} \times \mathbb{Z}$  are the two exam scores;
- $C_{90}(x, y) = (x + y)/2$ ;
- $E_{90} = \top$  if  $C_{90} > 90$  and  $E_{90} = \perp$  otherwise.

Similarly, we pack the checks for *Interview* and *EA* into CUs  $u_{int}$  and  $u_{ea}$ , resp, as follows.

$u_{int} = \langle T_{int}, C_{int}, E_{int} \rangle$  in which:

- $T_{int} = \{A, B, C, D, E\}$ ;
- $C_{int}(x) = x$ ;
- $E_{int} = \top$  if  $C_{int} = A$  and  $E_{int} = \perp$  otherwise.

$u_{ea} = \langle T_{ea}, C_{ea}, E_{ea} \rangle$  in which:

- $T_{ea} = \{\text{Yes}, \text{No}\}$ ;
- $C_{ea}(x) = x$ ;
- $E_{ea} = \top$  if  $C_{ea} = \text{Yes}$  and  $E_{ea} = \perp$  otherwise.

We use the following framework to model the admission problem.

- $\mathcal{U}$  is the following CUs:  
 $u_{90}(s_1)$   $u_{ea}(s_1)$   $u_{int}(s_1)$   
 $u_{90}(s_2)$   $u_{ea}(s_2)$   $u_{int}(s_2)$
- $\mathcal{L}$  is the following sentences:  

C1(s1)	C2(s1)	Ave>90(s1)	EA(s1)
notC1(s1)	notC2(s1)	Adm(s1)	INT(s1)
C1(s2)	C2(s2)	Ave>90(s2)	EA(s2)
notC1(s2)	notC2(s2)	Adm(s2)	INT(s2)



## REFERENCES

- [1] P. Dung. On the acceptability of arguments and its fundamental role in nonmonotonic reasoning, logic programming and n-person games. *AIJ*, 77(2):321–357, 1995.
- [2] X. Fan, S. Liu, H. Zhang, C. Leung, and C. Miao. Explained activity recognition with computational assumption-based argumentation. In *Proc. ECAI*, 2016.
- [3] X. Fan and F. Toni. Decision making with assumption-based argumentation. In *Proc. TAFIA*, pages 127–142. Springer, 2013.
- [4] X. Fan, F. Toni, A. Mocanu, and M. Williams. Multi-agent decision making with assumption-based argumentation. In *Proc. AAMAS*, pages 533–540, 2014.
- [5] X. Fan, H. Zhang, C. Leung, and C. Miao. A first step towards explained activity recognition with computational abstract argumentation. In *Proc. IEEE MFI*, 2016.
- [6] S. Modgil, F. Toni, F. Bex, I. Bratko, C. Chesñevar, W. Dvořák, M. Falappa, X. Fan, S. Gaggl, A. García, M. González, T. Gordon, J. Leite, M. Možina, C. Reed, G. Simari, S. Szeider, P. Torroni, and S. Woltran. The added value of argumentation. In *Agreement Technologies*, volume 8, pages 357–403. Springer, 2013.
- [7] Y. Peng and J. Reggia. *Abductive Inference Models for Diagnostic Problem Solving*. Springer, 1990.
- [8] F. Toni. A tutorial on assumption-based argumentation. *Argument & Computation, Special Issue: Tutorials on Structured Argumentation*, 5(1):89–117, 2014.
- [9] K. P. Yoon and C.-L. Hwang. *Multiple Attribute Decision Making: An Introduction*. Sage Publications Inc, March 1995.