

# An Adjustment Function For Dealing With Similarities

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# Outline

- Introduction
- Adjustment Function
- Properties
- Related Work
- Conclusions and Perspectives

# Introduction

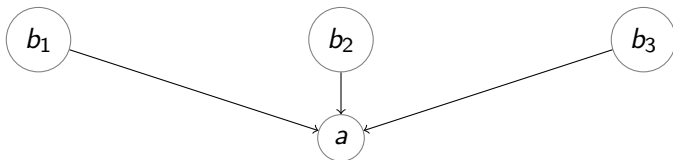
Argumentation is a **reasoning approach** based on **interacting arguments**

[a] My salary is quite high then I can rent the apartment.

[b<sub>1</sub>] If the bank approves the loan to buy a car, I will not have enough money to rent the apartment.

[b<sub>2</sub>] I may adopt a dog, however the owner forbids pets, thus I will not rent the apartment.

[b<sub>3</sub>] I may adopt a cat, however the owner forbids pets, thus I will not rent the apartment.



# Weighted Arguments

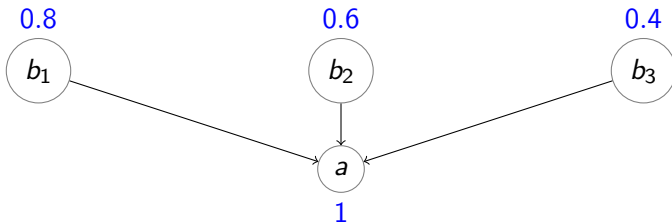
For instance: certainty degree

[a] My salary is quite high then I can rent the apartment.

[b<sub>1</sub>] If the bank approves the loan to buy a car, I will not have enough money to rent the apartment.

[b<sub>2</sub>] I may adopt a dog, however the owner forbids pets, thus I will not rent the apartment.

[b<sub>3</sub>] I may adopt a cat, however the owner forbids pets, thus I will not rent the apartment.



# Weighted Relations

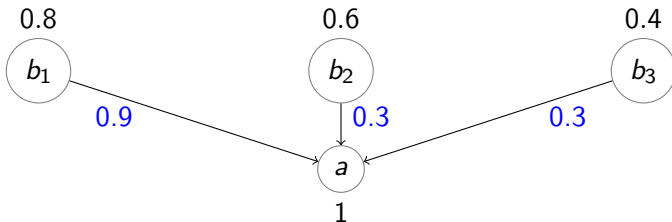
For instance: relevance degree

[a] My salary is quite high then I can rent the apartment.

[b<sub>1</sub>] If the bank approves the loan to buy a car, I will not have enough money to rent the apartment.

[b<sub>2</sub>] I may adopt a dog, however the owner forbids pets, thus I will not rent the apartment.

[b<sub>3</sub>] I may adopt a cat, however the owner forbids pets, thus I will not rent the apartment.



# Similarity between Arguments

[ $b_1$ ] If the bank approves the loan to buy a car, I will not have enough money to rent the apartment.

[ $b_2$ ] I may adopt a dog, however the owner forbids pets, thus I will not rent the apartment.

[ $b_3$ ] I may adopt a cat, however the owner forbids pets, thus I will not rent the apartment.

Similarity between arguments:

$$s(b_1, b_2) = s(b_1, b_3) = 0.2$$

$$s(b_2, b_3) = 0.9$$

# Argumentation Framework

## Definition (AF)

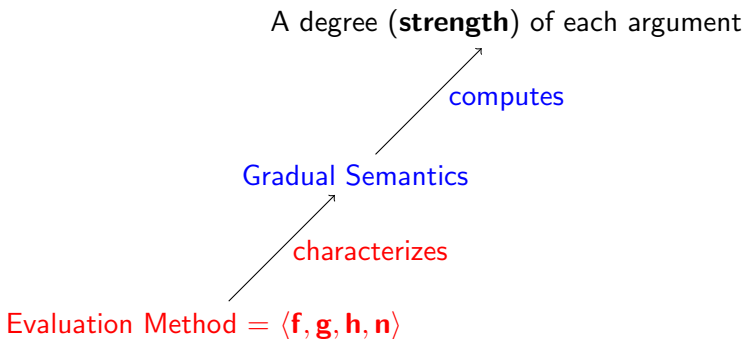
An argumentation framework (AF) is a tuple  $\mathbf{G} = \langle \mathcal{A}, \mathbf{w}, \mathcal{R}, \sigma, \mathbf{s} \rangle$ , where

- $\mathcal{A} \subseteq_f \text{Args}^a$
- $\mathbf{w} : \mathcal{A} \rightarrow [0, 1]$
- $\mathcal{R} \subseteq \mathcal{A} \times \mathcal{A}$  (Attack relation)
- $\sigma : \mathcal{R} \rightarrow [0, 1]$
- $\mathbf{s} : \mathcal{A} \times \mathcal{A} \rightarrow [0, 1]$  (Similarity measure)

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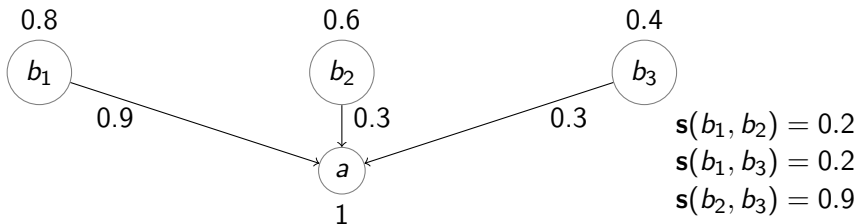
<sup>a</sup>Args denotes the universe of all possible arguments

# Evaluation of Argument Strength





# Evaluation Method $\mathcal{M} = \langle \mathbf{f}, \mathbf{g}, \mathbf{h}, \mathbf{n} \rangle$



- 1 Assess the **strenght** of every attack  $(b_i, a)$ ,  
 $\alpha_1 = \mathbf{h}(0.8, 0.9)$ ,  $\alpha_2 = \mathbf{h}(0.6, 0.3)$ ,  $\alpha_3 = \mathbf{h}(0.4, 0.3)$
- 2 **Adjust** the strenght of every attack **w.r.t. similarity**,  
 $\beta = \mathbf{n}((\alpha_1, b_1), (\alpha_2, b_2), (\alpha_3, b_3)) = (\beta_1, \beta_2, \beta_3)$
- 3 Assess the **strength** of the group of attacks on  $a$ ,  
 $\delta = \mathbf{g}(\beta_1, \beta_2, \beta_3)$
- 4 Evaluate the **impact** of attacks on the initial weight of  $a$ ,  
 $\lambda = \mathbf{f}(1, \delta)$

# Evaluation Method $\mathcal{M} = \langle \mathbf{f}, \mathbf{g}, \mathbf{h}, \mathbf{n} \rangle$

## Definition (EM)

An *evaluation method* (EM) is a tuple  $\mathcal{M} = \langle \mathbf{f}, \mathbf{g}, \mathbf{h}, \mathbf{n} \rangle$  such that:

- $\mathbf{f} : [0, 1] \times \text{Range}(\mathbf{g})^a \rightarrow [0, 1]$ ,
- $\mathbf{g} : \bigcup_{k=0}^{+\infty} [0, 1]^k \rightarrow [0, +\infty[$ ,
- $\mathbf{h} : [0, 1] \times [0, 1] \rightarrow [0, 1]$ ,
- $\mathbf{n} : \bigcup_{k=0}^{+\infty} ([0, 1] \times \text{Args})^k \rightarrow [0, 1]^k$ .

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<sup>a</sup>Range( $\mathbf{g}$ ) denotes the co-domain of  $\mathbf{g}$

# Gradual Semantics

## Definition (Gradual Semantics - $\mathcal{S}$ )

$\mathcal{S}$  based on an evaluation method  $\mathcal{M} = \langle \mathbf{f}, \mathbf{g}, \mathbf{h}, \mathbf{n} \rangle$  is a function assigning to every AF  $\mathbf{G} = \langle \mathcal{A}, \mathbf{w}, \mathcal{R}, \sigma, \mathbf{s} \rangle$ , a weighting  $\text{Deg}_{\mathbf{G}}^{\mathcal{S}} : \mathcal{A} \rightarrow [0, 1]$  such that  $\forall a \in \mathcal{A}$ ,

$$\text{Deg}_{\mathbf{G}}^{\mathcal{S}}(a) = \mathbf{f} \left( \mathbf{w}(a), \mathbf{g} \left( \mathbf{n} \left( \left( \mathbf{h}(\text{Deg}_{\mathbf{G}}^{\mathcal{S}}(b_1), \sigma(b_1, a)), b_1 \right), \dots, \right. \right. \right. \\ \left. \left. \left. \left( \mathbf{h}(\text{Deg}_{\mathbf{G}}^{\mathcal{S}}(b_k), \sigma(b_k, a)), b_k \right) \right) \right) \right),$$

where  $\{b_1, \dots, b_k\} = \text{Att}(a)^a$ .

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<sup>a</sup>Att( $a$ ) denotes the set of attackers of  $a$

$\text{Deg}_{\mathbf{G}}^{\mathcal{S}}(a)$  represents the strength of  $a$

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# Weighted h-Categorizer

## Definition (Weighted h-Categorizer)

The function  $\mathcal{S}_{wh}$  transforms any AF  $\mathbf{G} = \langle \mathcal{A}, \mathbf{w}, \mathcal{R}, \sigma, \mathbf{s} \equiv \mathbf{0} \rangle$ , into a weighting  $\text{Deg}_{\mathbf{G}}^{\mathcal{S}_{wh}} : \mathcal{A} \rightarrow [0, 1]$  s.t.  $\forall a \in \mathcal{A}$ ,

$$\text{Deg}_{\mathbf{G}}^{\mathcal{S}_{wh}}(a) = \begin{cases} \mathbf{w}(a) & \text{iff } \text{Att}(a) = \emptyset \\ \frac{\mathbf{w}(a)}{1 + \sum_{b \in \text{Att}(a)} \text{Deg}_{\mathbf{G}}^{\mathcal{S}_{wh}}(b) \times \sigma(b, a)} & \text{else} \end{cases}$$

$\mathcal{S}_{wh}$  is based on  $\mathcal{M} = \langle \mathbf{f}_{\text{frac}}, \mathbf{g}_{\text{sum}}, \mathbf{h}_{\text{prod}} \rangle$  such that:

$$\begin{cases} \mathbf{f}_{\text{frac}}(x_1, x_2) = \frac{x_1}{1+x_2} \\ \mathbf{g}_{\text{sum}}(x_1, \dots, x_n) = \sum_{i=1}^n x_i \\ \mathbf{h}_{\text{prod}}(x_1, x_2) = x_1 \times x_2 \end{cases}$$

Adjustment function  $\mathbf{n}_{\text{wh}}$ Definition ( $\mathbf{n}_{\text{wh}}$ )

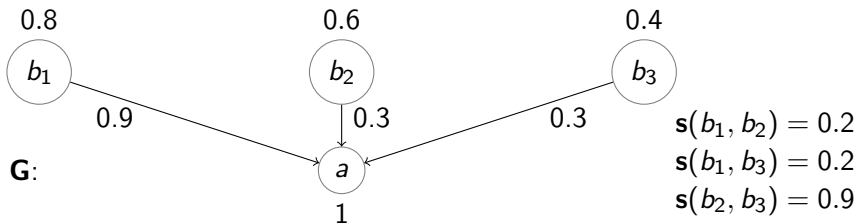
Let  $\mathbf{G} = \langle \mathcal{A}, \mathbf{w}, \mathcal{R}, \sigma, \mathbf{s} \rangle$  be an AF,  $x_1, \dots, x_k \in [0, 1]$ , and  $b_1, \dots, b_k \in \mathcal{A}$ . We define  $\mathbf{n}_{\text{wh}}$  as follows:

$$\mathbf{n}_{\text{wh}}((x_1, b_1), \dots, (x_k, b_k)) = (\text{Deg}_{\mathbf{G}'}^{S_{\text{wh}}}(b_1), \dots, \text{Deg}_{\mathbf{G}'}^{S_{\text{wh}}}(b_k))$$

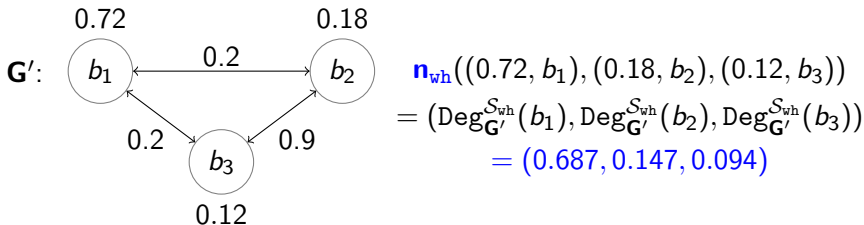
where  $\mathbf{G}' = \langle \mathcal{A}', \mathbf{w}', \mathcal{R}', \sigma', \mathbf{s}' \rangle$ , such that:

- $\mathcal{A}' = \{b_1, \dots, b_k\}$ ,
- $\mathbf{w}'(b_1) = x_1, \dots, \mathbf{w}'(b_k) = x_k$ ,
- $\mathcal{R}' = \{(b_1, b_2), \dots, (b_1, b_k), \dots, (b_k, b_1), \dots, (b_k, b_{k-1})\}$ ,
- $\forall (b_i, b_j) \in \mathcal{R}', \sigma'(b_i, b_j) = \mathbf{s}(b_i, b_j)$ ,
- $\mathbf{s}' \equiv 0$ .

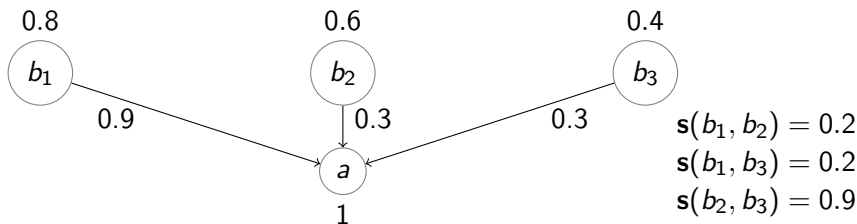
# Example of $n_{wh}$



Assume that we use  $\mathbf{h}_{\text{prod}}(x_1, x_2) = x_1 \times x_2$



# Example of $EM_{wh} = \langle \mathbf{f}_{frac}, \mathbf{g}_{sum}, \mathbf{h}_{prod}, \mathbf{n}_{wh} \rangle$



- Strength of every attack,  $\alpha_1 = \mathbf{h}_{prod}(0.8, 0.9) = 0.72$ ,  
 $\alpha_2 = \mathbf{h}_{prod}(0.6, 0.3) = 0.18$ ,  $\alpha_3 = \mathbf{h}_{prod}(0.4, 0.3) = 0.12$
- Strength of every attack w.r.t. similarity,  $(\beta_1, \beta_2, \beta_3) =$   
 $\mathbf{n}_{wh}((\alpha_1, b_1), (\alpha_2, b_2), (\alpha_3, b_3)) = (0.687, 0.147, 0.094)$
- Strength of the group of attacks on  $a$ ,  
 $\delta = \mathbf{g}_{sum}(\beta_1, \beta_2, \beta_3) = 0.928$
- Strength of  $a$ ,  $\lambda = \mathbf{f}_{frac}(1, \delta) = 0.519$



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# Some Properties on $\mathbf{n}_{wh}$

$\mathbf{n}_{wh}$  can be used by a gradual semantics

## Theorem

*There exists a unique semantics that is based on the evaluation method  $\langle \mathbf{f}_{frac}, \mathbf{g}_{sum}, \mathbf{h}_{prod}, \mathbf{n}_{wh} \rangle$ .*

## Some Properties of $\mathbf{n}_{\text{wh}}$

When all the arguments are **dissimilar**, the adjustment function **doesn't alter their initial values**

### Property (P1)

For any AF  $\mathbf{G} = \langle \mathcal{A}, \mathbf{w}, \mathcal{R}, \sigma, \mathbf{s} \rangle$ , for all  $a_1, \dots, a_n \in \mathcal{A}$ , for all  $x_1, \dots, x_n \in [0, 1]$ , if  $\forall i, j \in \{1, \dots, n\}, i \neq j, \mathbf{s}(a_i, a_j) = 0$ , then

$$\mathbf{n}_{\text{wh}}((x_1, a_1), \dots, (x_n, a_n)) = (x_1, \dots, x_n).$$

$\mathbf{n}_{\text{wh}}$  can **only reduce the value** of an argument

### Property (P2)

For any AF  $\mathbf{G} = \langle \mathcal{A}, \mathbf{w}, \mathcal{R}, \sigma, \mathbf{s} \rangle$ , for all  $a_1, \dots, a_n \in \mathcal{A}$ , for all  $x_1, \dots, x_n \in [0, 1]$ , if  $\mathbf{n}_{\text{wh}}((x_1, a_1), \dots, (x_n, a_n)) = (x'_1, \dots, x'_n)$ , then  $\forall i \in \{1, \dots, n\}, x'_i \leq x_i$ .

## Some Properties of $\mathbf{n}_{wh}$

If an argument is **dissimilar** to all other arguments **and** its initial **value is 0**, then it will **not have any impact** on the readjusted values of the other arguments

### Property (P3)

For any AF  $\mathbf{G} = \langle \mathcal{A}, \mathbf{w}, \mathcal{R}, \sigma, \mathbf{s} \rangle$ , for all  $a_1, \dots, a_n, b \in \mathcal{A}$ , for all  $x_1, \dots, x_n, y \in [0, 1]$ , if

- $\forall i \in \{1, \dots, n\}, \mathbf{s}(a_i, b) = 0,$
- $y = 0,$

then

$$\mathbf{n}_{wh}((x_1, a_1), \dots, (x_n, a_n), (y, b)) = (\mathbf{n}_{wh}((x_1, a_1), \dots, (x_n, a_n)), 0).$$

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# Readjusted Score - [Amgoud and al., KR 2018]

avg denotes the average operator

## Definition ( Readjusted Score $\mathbf{n}_{rs}$ )

Let  $a_1, \dots, a_k \in \text{Args}$  and  $x_1, \dots, x_k \in [0, 1]$ .

$\mathbf{n}_{rs}((x_1, a_1), \dots, (x_k, a_k)) =$

$$\left( \underset{x_i \in \{x_1, \dots, x_k\} \setminus \{x_1\}}{\text{avg}} \left( \frac{\text{avg}(x_1, x_i) \times (2 - \mathbf{s}(a_1, a_i))}{2} \right), \dots, \underset{x_i \in \{x_1, \dots, x_k\} \setminus \{x_k\}}{\text{avg}} \left( \frac{\text{avg}(x_k, x_i) \times (2 - \mathbf{s}(a_k, a_i))}{2} \right) \right).$$

$\mathbf{n}_{rs}() = ()$  and  $\mathbf{n}_{rs}((x_1, a_1)) = (x_1)$  if  $k = 1$ .

# Properties on $\mathbf{n}_{rs}$

## Property

- 1  $\mathbf{n}_{rs}$  violates *P1*, *P2* and *P3*
- 2 Let  $\mathbf{g}_{\text{sum}}(x_1, \dots, x_n) = \sum_{i=1}^n x_i$ ,  $\exists a_1, \dots, a_n, b \in \text{Args}$  and  $x_1, \dots, x_n, y \in [0, 1]$  such that:
  - $\forall i \in \{1, \dots, n\}, \mathbf{s}(a_i, b) = 0$ ,
  - $y = 0$ ,
  - $\mathbf{g}_{\text{sum}}(\mathbf{n}_{rs}((x_1, a_1), \dots, (x_n, a_n))) < \mathbf{g}_{\text{sum}}(\mathbf{n}_{rs}((x_1, a_1), \dots, (x_n, a_n), (y, b)))$

# Conclusions and Perspectives

## Conclusions:

- Extending the notion of evaluation method by an **adjustment function**
  - Proposing a **novel adjustment function  $n_{wh}$**
  - Investigating the **properties** of the existing functions
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## Perspectives:

- **Axiomatisation** of adjustment functions
- **Investigating** other adjustment functions



## Reference

**[Amgoud and al., KR 2018]**: Amgoud L, Bonzon E, Delobelle J, Doder D, Konieczny S, Maudet N. Gradual Semantics Accounting for Similarity between Arguments. In: KR; 2018. p. 88–97.