Normalized Model Evaluation Concerning Social Abstract Argumentation

Sinan Eğilmez Supervised by: João Leite

February 17, 2015





Outline

- 1. Social Abstract Argumentation Frameworks
- 2. Normalized Model Evaluation Concerning SAF

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Motivation

- Interactions in Social Networks are unstructured, often chaotic.
- Prevents a fulfilling experience for those seeking deeper interactions and not just increasing their number of friends, likes, etc.

The Vision: A self-managing online debating system capable of accommodating two archetypal levels of participation:

- experts/enthusiasts who specify arguments and the attacks between arguments.
- observers/random browsers will vote on individual arguments, and on the specified attacks via GUI.
- ▶ autonomously maintaining a formal outcome to debates by assigning a strength to each argument based on the structure of the argumentation graph and the votes

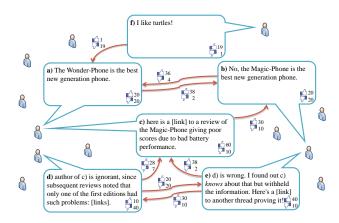
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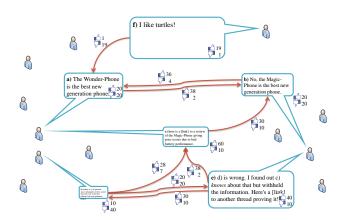
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Envisioned Framework



Envisioned Framework Model



- $F = \langle A, \mathcal{R}, V_A, V_{\mathcal{R}} \rangle$
- $\blacktriangleright S = \langle L, A, A, A, \Upsilon, \neg, \tau \rangle$
- Desiderata for Semantics
- ► Class of Well-behaved semantics
- ▶ Concrete Semantics, $S_{\epsilon} = \langle [0,1], \mathcal{L}, \mathcal{L},$
- ▶ A S-model of F is a total mapping $M : A \rightarrow L$ such that for all $a \in A$

$$M(a) = \tau(a) \curlywedge_{\mathcal{A}} \neg \bigvee_{a_i \in \mathcal{R}^-(a)} (\tau((a_i, a)) \curlywedge_{\mathcal{R}} M(a_i))$$

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 - ▶ Social voting on arguments [Leite11]
 - Social voting on attack relations [Egilmez13]
- ► Flexible argument valuation
- Enjoys desirable properties (under Product semantics)
 - e.g. Fairness, Existence and Uniqueness of Models
- Fast implementations of model evaluation [Correia14]
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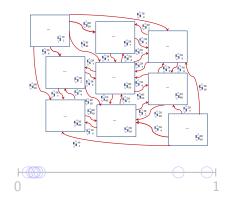
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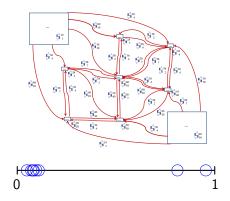
Problem motivation

▶ When the social argumentation graph is dense, the model evaluations tend to get smaller.



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Formal Roadmap

- ► Characterization of a normalized dataset
 - Concepts from the field of Statistics
- Desired properties/mappings
 - Existence of arguments
 - ► Relative ordering of model evaluations
 - Relative ordering of distances
 - Upper limit by social support
- Construction of the normalizing algorithm
 - In alignment with classes of desired mappings

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 - Cluster density
 - Time/Space complexity
 - Clustering spacing
- 2. Normalizing phase (via update function)



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- 1. Labeling phase (by clustering)
 - Cluster density
 - Time/Space complexity
 - Clustering spacing
- 2. Normalizing phase (via update function)
 - Desirable properties
 - Time complexity



Ongoing Work

- Automated means for characterization.
 - ▶ Initial attempts rely on expert knowledge or constants.
- Investigation of desirable properties, stricter desirable mapping classes.
- Investigation of algorithms.

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- Aho, Hopcroft, and Ullman, *The Design and Analysis of Computer Algorithms*, 1974

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Bibliography

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- M. Correia, J. Cruz and J. Leite, On the Efficient Implementation of Social Abstract Argumentation, In Procs. of ECAI 2014.



```
input : D (instances set), k (# of clusters))
   output: D (normalized instances set)
 1 cluster\_counter = 0;
 2 Clusters ← ∅;
 3 foreach d_i \in \mathcal{D} do
       newCluster \leftarrow \{d_i\};
       instance\_counter(newCluster) = 1;
       Clusters \leftarrow Clusters \cup newCluster;
       cluster\_counter + +:
 s end
 9 foreach C_i, C_i \in Clusters do
       Compute \Lambda(C_i, C_i)
11 end
12 while cluster_counter \neq k do
       (C_m, C_n) = two clusters closest together in Clusters;
13
       Clusters \leftarrow Clusters -\{C_m\} - \{\bar{C}_n\} \cup \{C_m \cup C_n\};
14
       cluster\_counter - -:
15
       foreach C \in Clusters do
16
           Compute \bigwedge(C, (C_m \cup C_n));
18
       end
19 end
20 Mid ← ∅;
21 foreach C_i, C_{i+1} \in Clusters do
       Mid \leftarrow Mid \cup \{(max(C_i) + min(C_{i+1})/2)\};
23 end
24 foreach C_i \in Clusters \backslash C_k do
       foreach d \in C_i do
25
           d \leftarrow \frac{d - min(C_i)}{max(C_i) - min(C_i)} (Mid[i] - min(C_i)) + min(C_i);
26
       end
27
28 end
```