

University of Tsukuba, LGS'09, 26 - 29 August 2009

White Manipulation in Judgment Aggregation



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What is this all about

- judgment aggregation (JA) has two problems:
 - aggregation functions that satisfy a desirable set of properties do not exist
 - aggregation operators that exist are manipulable
- the question is: is lying, cheating and manipulation really that bad ?

the colloquial term "white lies"

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Wiktionary	Registration for Wikimania 2009 is now open. Learn more.	[Hide] [Help us with translations!]
['wIk∫ənrI] n., a wiki-based Open Content dictionary	News for editors; Propose a logo for Wiktionary white lie	
Wilco ['wrl kars] search	English	[edit]
Go Search	Noun	[edit]
navigation	white lie (plural white lies) Wikiped	dia has an article on:
Main Page Community portal	1. (<i>idiomatic</i>) A deliberate, untrue statement which does no harm or is intended to produce a favorable result.	e lie
= Wiktionary	= 2008, Jacqueline Stenson, "The Whole Truth: When is it okay to lie to your kids? @," Newsweek, 15 Jul.,	
 preferences Requested entries Recent changes 	An occasional little white lie such as Weston's probably won't cause any lasting damage. And at times, tell the whole truth to a child who's not at an age to handle it—may do more harm than good, they say.	ling the truth—particularly
Random entry	Translations	[edit]

3

the colloquial term "white lies"

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manipulation - lying with the intent to improve the outcome for the agent who lies

the colloquial term "white lies"

- manipulation lying with the intent to improve the outcome for the agent who lies
- white manipulation lying with the intent to improve the outcome for all the agents involved

In the rest of the talk

- introduce the basic concepts of judgment aggregation
- redefine the judgment aggregation function
- Introduce in JA: scoring functions, social welfare notions
- define white manipulation
- initial results
- conclusions

- how individual judgments on logically connected issues can be aggregated into a collective judgment on the same issues
- hiring committee example with rule $x \leftrightarrow (a \land b)$

	a = X is good at teaching	b = X is good at research	x = hire X
prof. A	yes	no	no
prof. B	yes	yes	yes
prof. C	no	yes	no
Majority	yes	yes	no

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impasse

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judgment aggregation functions are not manipulable if they satisfy independence and (weak) monotonicity[1]

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prof. B	ye	25	ye	S	yes
prof. C	n	0	$y \epsilon$	S	no
Majority	ye	28	$y\epsilon$	28	no

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prof. B	y_{ϵ}	2.5	ye	S	y	cs
prof. C	n	0	ye	s	n	0
Majority	$y\epsilon$	28	ye	25	n	0

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the premise based procedure

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the premise based procedure is manipulable [2]

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the conclusion based procedure

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the conclusion based procedure is manipulable [3]

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distance based merging

Judgment Aggregation

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distance based merging is manipulable [4]

The ideas are ...

- impasse ↓ is in the set of possible outputs of the aggregation function, but not part of any profile
- assume that agents have preferences over outputs and neither of the agents prefers the output ↓
- scoring functions determine preference ordering over the elements of the set of possible outputs of the aggregation function

- JA function we defined as $f: \Omega \longrightarrow \Phi \cup \{\downarrow\}$
 - example we work with quota rule f^q
- score function we define as a function that, given a judgment set, scores all other possible outcomes based on that judgment set
- we work with examples of distance based scoring functions: $HS_e: \Phi \longrightarrow (\Phi^{\downarrow} \longrightarrow \mathbb{N})$
 - $VS_k: \Phi \longrightarrow (\Phi^{\downarrow} \longrightarrow \mathbb{N})$

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Definition. [Manipulation] Let s be a scoring function. An aggregation function f is manipulable if and only if there exists a judgment profile $\omega \in \Omega$ and an agent isuch that $f(\omega) \prec_i^s f(\omega')$, where $\omega' \in \Omega$ is some *i*-variant of ω .

 $VS_k: \Phi \longrightarrow (\Phi^{\downarrow} \longrightarrow \mathbb{N})$

Social welfare notions in JA

- using a scoring function, a preference profile can be built from a judgment profile
- having a preference profile, social welfare notions can be applied to JA
- Utilitarian social welfare $USW_{s(\omega)}(\varphi) = \sum_{i=1}^{n} s(\varphi)(\varphi_i)$

Egalitarian social welfare

 $ESW_{s(\omega)}(\varphi) = \max\{s(\varphi)(\varphi_i) \mid \varphi_i \in \Phi\}$

Social welfare notions in JA

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Social welfare notions in JA

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Definition. [White manipulability] Let \mathcal{SW} be a social welfare function and s a scoring function. An aggregation function fis white manipulable if and only if there exists an agent i and a judgment profile $\omega \in \Omega$ such that $f(\omega) \prec_i^s f(\omega')$ and $\mathcal{SW}(f(\omega)) < \mathcal{SW}(f(\omega'))$, where $\omega' \in \Omega$ is some *i*-variant of ω .

Φ	(a,b,x)	$x \leftrightarrow (a \land b)$
(prof. A) φ_1	$(1,\!0,\!0)$	
(prof. B) φ_2	(1,1,1)	
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20 possible profiles

Φ	(a,b,x)	$x \leftrightarrow (a \land b)$
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$arphi_4$	$(0,\!0,\!0)$	

 $(1,0,0) \succ_{2}^{HS_{e}} (0,0,0) \succ_{2}^{HS_{e}} (0,1,0) \sim_{2}^{HS_{e}} (1,1,1) \succ_{2}^{HS_{e}} \downarrow$ $(1,1,1) \succ_{1}^{HS_{e}} (1,0,0) \sim_{1}^{HS_{e}} (0,1,0) \succ_{1}^{HS_{e}} (0,0,0) \succ_{1}^{HS_{e}} \downarrow$ $(0,1,0) \succ_{3}^{HS_{e}} (0,0,0) \succ_{3}^{HS_{e}} (1,0,0) \sim_{3}^{HS_{e}} (1,1,1) \succ_{3}^{HS_{e}} \downarrow$

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 $f(\omega)$

Coordinated white manipulation

- one agent can white manipulate alone and improve the social welfare
- the group can agree on how to manipulate and this improve the social welfare
- idea: negotiate on how to lie
- example: fallback bargaining

- introduced by S.J. Brams and D.M. Kilgour (1998)[5]
- bargainers "fallback" on less and less preferred alternatives

$$M = \begin{pmatrix} a & b & c & d \\ a & c & b & d \\ b & a & d & c \end{pmatrix}_{d=1 \ d=2 \ d=3 \ d=4}$$

hiring example:

$$M^{h} = \begin{pmatrix} 100 & 000 & 111,010 \downarrow \\ 111 & 100,010 & 000 \downarrow \\ 010 & 000 & 111,100 \downarrow \\ d=1 & d=2 & d=3 & d=4 \end{pmatrix}$$

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$$M^{h} = \begin{pmatrix} 100 & 000 & 111,010 \downarrow \\ 111 & 100,010 & 000 \downarrow \\ 010 & 000 & 111,100 \downarrow \\ d = 1 & d = 2 & d = 3 & d = 4 \end{pmatrix} \mathbf{r} = \mathbf{n}$$

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- bargainers "fallback" on less and less preferred alternatives

$$M = \begin{pmatrix} a & b & c & d \\ a & c & b & d \\ b & a & d & c \end{pmatrix}_{d=1 \ d=2 \ d=3 \ d=4}$$

hiring example:

$$M^{h} = \begin{pmatrix} 100 & 000 & 111,010 | \downarrow \\ 111 & 100,010 & 000 & \downarrow \\ 010 & 000 & 111,100 | \downarrow \\ d = 1 & d = 2 & d = 3 & d = 4 \end{pmatrix} \mathbf{r} = \mathbf{n}$$

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hiring example:

$$M^{h} = \begin{pmatrix} 100 & 000 & 111,010 \downarrow \\ 111 & 100,010 & 000 \downarrow \\ 010 & 000 & 111,100 \downarrow \\ d = 1 & d = 2 & d = 3 & d = 4 \end{pmatrix} \mathbf{r} = \mathbf{2}$$

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- bargainers "fallback" on less and less preferred alternatives

$$M = \begin{pmatrix} a & b & c & d \\ a & c & b & d \\ b & a & d & c \end{pmatrix}_{d=1 \ d=2 \ d=3 \ d=4}$$

hiring example:

$$M^{h} = \begin{pmatrix} 100 & 000 & 111,010 \downarrow \\ 111 & 100,010 & 000 & \downarrow \\ 010 & 000 & 111,100 \downarrow \\ d = 1 & d = 2 & d = 3 & d = 4 \end{pmatrix} \mathbf{r} = \mathbf{2}$$

Fallback Bargaining & WM

- if ↓ is the least preferred outcome, it will not be the result of the bargaining (for both scoring functions)
- for r=n, the utilitarian social welfare of the bargaining output is the highest

Our contribution

- treat the inconsistency as an impasse and the impasse as a possible outcome
- introduce the idea of manipulability as a positive concept
- extend the judgment aggregation framework with an automatically built preference profile
- Introduce social welfare concepts in the judgment aggregation framework

Future work

- analyze further the fallback bargaining for other social welfare functions
- analyze other agreement reaching protocols for the use of white manipulation
- analyze profiles with different preferences regarding the impasse
- redefine manipulation concepts in terms of coalition manipulation concepts
- extend the JA framework towards game theory

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