### Crowdsourcing Pareto-Optimal Object Finding by Pairwise Comparisons

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# Humans' Obsession: Comparing Things

### Who is the better footballer?



http://visual.ly/messi-vs-ronaldo

### Which is the better company to work for?

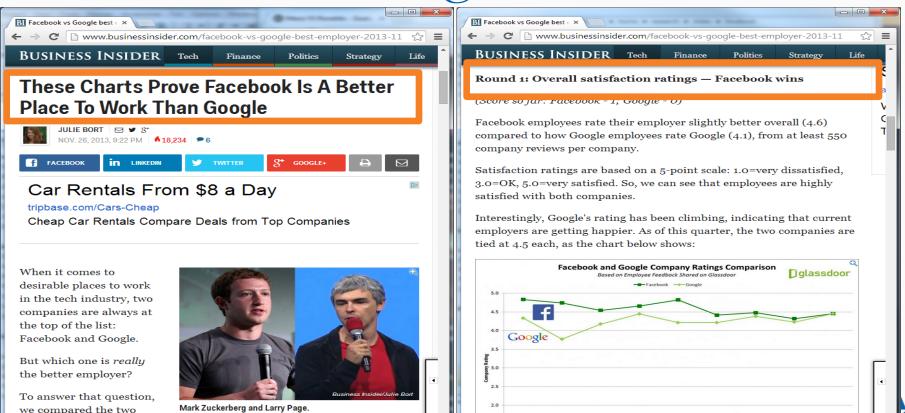






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# Facebook vs. Google



http://www.businessinsider.com/facebook-vs-google-best-employer-2013-11 ©2015 The University of Texas at Arlington. All Rights Reserved.

# Facebook vs. Google

#### Multiple criteria

1				1
Overall satisfaction	CEO approval	Employee confidence in the future	Perks and salaries	Interview difficulty
<b>F</b> > <b>8</b>	<b>F3</b> > <b>8</b>	<b>F3</b> > <b>8</b>	<u>8</u> > F	FI > 8
	·	γ	·	J
	Which	n one is be	tter?	
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### Overview

- 1. Ask crowd to compare objects on individual criteria
- 2. Derive partial knowledge about preference relations based on responses from crowd
- 3. Find all Pareto-optimal objects without exhausting all possible comparison questions

General framework and its instantiations (algorithms), with the goal of minimizing number of questions

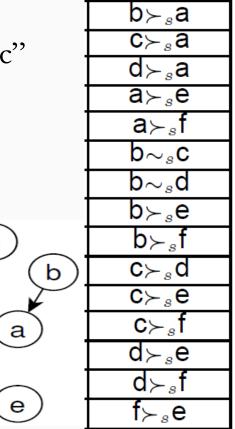
### Preference (Better-Than) Relation $P_c$ $(x, y) \in P_c \text{ (or } x \succ_c y)$ "x is better than (preferred over) y with regard to criterion c"

Assumptions on data model:

- P<sub>c</sub> is a strict partial order (as opposed to a total order) Irreflexivity, Transitivity, Asymmetry
- No explicit attribute representation, thus no equivalence on a criterion

#### Notations:

- o (indifferent) x ∼<sub>c</sub> y ⇔ (x, y) ∉ P<sub>c</sub> ∧ (y, x) ∉ P<sub>c</sub>
- $\circ \quad (\text{not better than}) \ x \not\succ_c y \ \Leftrightarrow \ (x, y) \not\in P_c$



d

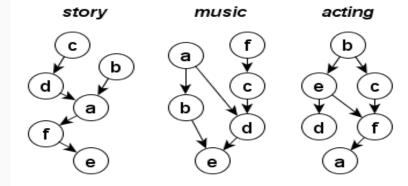
## Pareto-optimal Objects

Object dominance: consider objects O, criteria C  $y > x \iff \forall c \in C : x \succeq_c y \text{ and } \exists c \in C \text{ such that } y \succeq_c x \text{ (i.e., x is not better than y by any criterion and y is better than x by at least one criterion)$ 

 $x \in O$  is Pareto-optimal  $\Leftrightarrow x$  is not dominated by any other object

Example:

- $\circ$  c > d  $\Leftarrow$  c ><sub>story</sub> d, c ><sub>music</sub> d, c ~<sub>acting</sub> d
- o Only one Pareto-optimal object: b



### Deriving Preference Relations by Aggregating Crowd's Responses to Pairwise Comparisons

Between A separation(2011) and The big Lebowski(1998) which movie is better with regard to story?

- A separation(2011).
- The big Lebowski(1998).
- no preference.

submit skip this question

In each following pair(each row), which picture has more interesting colors?

Left Picture	Right Picture	Preference
		<ul> <li>Left Picture</li> <li>Right Picture</li> <li>No Preference</li> </ul>

Pareto-Optimal Object Finding

**Problem statement:** Given objects O and criteria C, find all Pareto-optimal objects, using pairwise comparisons by individual criteria

Cost metric

- o Goal: as few pairwise comparison questions as possible
- o Simple, but reflect real-world monetary cost and time delay
- o Brute-force approach :  $|C| \ge |O| \ge (|O| 1)/2$  questions

#### Assumptions on execution model

- o Sequential execution: get  $rlt(q_i)$  before asking  $q_{i+1}$
- No consideration of worker quality



# Applications

#### Collecting Public Opinion

Best companies to work for, best cities to live in
 Group Decision Making

• Where for lunch, which product to use, which candidate to hire Information Exploration

• Compare photos by color, sharpness, and landscape Back to the "which one is better"?

After finding Pareto-optimal objects, further actions (ranking, filtering, visualization) to find desirable objects

## Related Work

	Task	Question type	Multiple	Order among objects	Explicit attribute
			attributes	(on each attribute)	representation
[12]	full ranking	pairwise comparison	no	total order	no
[25]	top- <i>k</i> ranking	rank subsets of objects	no	total order	no
[15]	top-k ranking and grouping	pairwise comparison	no	total order	no
[23]	skyline/Pareto-optimal queries	missing value inquiry	yes	total order	yes
[17]	skyline/Pareto-optimal queries	pairwise comparison	yes	total order	no
[4], this work	skyline/Pareto-optimal queries	pairwise comparison	yes	strict partial order	no

[12] Chen et al. WSDM13 || [15] Davidson et al. ICDT13 || [17] Grozet al. PODS15 ||
[23] Lofi et al. EDBT13 || [25] Polychronopoulos et al. WebDB13 || [4] technical report of this paper

Other related work: collaborative filtering, learning to rank, ...



# Related Work

#### Explicit attribute representation

- Total order on ordinal attributes (sizes of houses, ratings of restaurants), Partial-order on categorical attributes (genres of movies)
- Not always easy to model and/or for users to provide missing values (e.g., story of movies)

#### Pairwise comparisons

• Known to be easier, faster, and less error-prone. Widely used in social choice and welfare, preferences, and voting

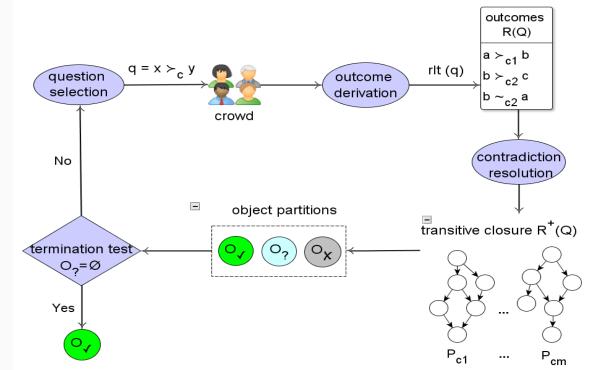
#### Partial order vs. Total order

- o A direct effect of using pairwise comparisons
- o Not always natural to enforce a total order



### General, Iterative Algorithm Framework

- 4-steps in each iteration
- (1) Question selection
- (2) Outcome derivation
- (3) Contradiction resolution
- (4) Termination test



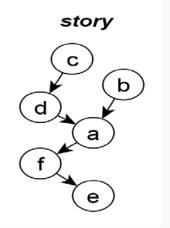
(2) Outcome Derivation

The framework is agnostic to the outcome derivation method. Other conceivable method can be plugged in.

Question

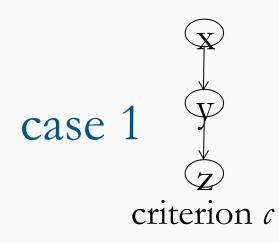
o  $q = x ?_c y$  (compare x and y by criterion c) Three possible outcomes based on voting

 $rlt(q) = \begin{cases} x \succ_{c} y \\ y \succ_{c} x \\ x \sim_{c} y \end{cases}$ 



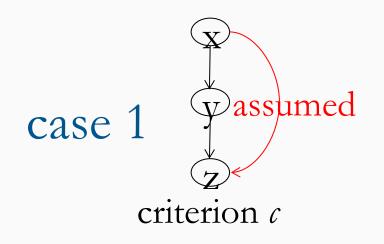
	ANSWER			
QUESTION	$\succ$	$\sim$	$\prec$	OUTCOME
a? <sub>s</sub> b	1	0	4	b≻₅a
a? <sub>s</sub> c	0	0	5	c≻₃a
a? <sub>s</sub> d	0	2	3	d≻₃a
a? <sub>s</sub> e	4	0	1	a≻₅e
a?₅f	3	1	1	a≻₅f
b?₅C	1	2	2	$b_{\sim_s}c$
b? <sub>s</sub> d	1	3	1	$b_{\sim_s}d$
b?se	5	0	0	b≻₅e
b? <sub>s</sub> f	4	1	0	b≻₅f
<b>c</b> ? <sub>s</sub> d	3	2	0	c≻₅d
c?se	4	0	1	c≻₅e
c? <sub>s</sub> f	3	1	1	c≻₅f
d?se	3	0	2	d≻₅e
d?₅f	3	2	0	$d_{\succ_s}f$
e? <sub>s</sub> f	1	1	3	f≻₅e

Assume transitivity in preference relation, and enforce it.



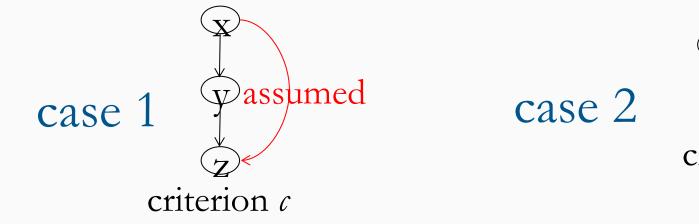


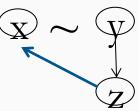
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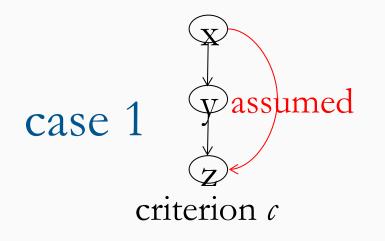


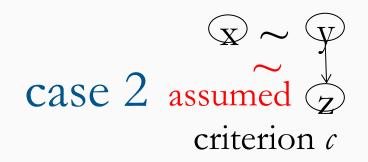


criterion c



Assume transitivity in preference relation, and enforce it.







# (4) Termination Test

At the end of each iteration, objects are partitioned into 3 sets, based on incomplete preference relations  $R^+(Q)$  so far.

$$\begin{split} &O_{\checkmark} = \{ \mathsf{x} \in O \mid \forall \mathsf{y} \in O : (\exists c \in C : \mathsf{x}\succ_c \mathsf{y} \in R^+(Q)) \lor (\forall c \in C : \mathsf{x}\succ_c \mathsf{y} \in R^+(Q)) \}; \\ &\mathsf{x}\sim_c \mathsf{y} \in R^+(Q)) \}; \\ &O_{\times} = \{ \mathsf{x} \in O \mid \exists \mathsf{y} \in O : (\forall c \in C : \mathsf{y}\succ_c \mathsf{x} \in R^+(Q) \lor \mathsf{x}\sim_c \mathsf{y} \in R^+(Q)) \land (\exists c \in C : \mathsf{y}\succ_c \mathsf{x} \in R^+(Q)) \}; \\ &\in R^+(Q)) \land (\exists c \in C : \mathsf{y}\succ_c \mathsf{x} \in R^+(Q)) \}; \\ &O_? = O \backslash (O_{\checkmark} \cup O_{\times}). \end{split}$$

- o OV: Pareto-optimal objects
- o OX: Non Pareto-optimal objects
- $\circ$  O? :  $R^+(Q)$  is insufficient for discerning these objects' Pareto-optimality
- $\circ$  O? =  $\varnothing \Rightarrow$  terminate



# (1) Question Selection

- The process of executing a question sequence  $Q = \langle q1, ..., qn \rangle$
- Q is a terminal sequence if  $O? = \emptyset$  based on  $R^+(Q)$ .
- Goal: among many terminal sequences, execute a short sequence
   Lower bound
- Theorem 2: At Ideast  $(|O|-k) \times |C| + (k-1) \times 2$  pairwise comparison questions are necessary, where k is the number of Pareto-optimal objects.

#### Bad news

- Worst-case :  $|C| \ge |O| \ge (|O|-1)/2$  questions; cannot do better than brute-force
- E.g., suppose all objects are indifferent by every criterion. If any comparison  $x ?_c y$  is skipped, we will not be able to determine if x and y are indifferent or if one dominates another.



Transitivity of Object Dominance: Doesn't Hold

- A cost-saving property for skyline queries
- Object dominance transitivity:  $x > y, y > z \Rightarrow x > z$
- Immediately prune a dominated object from further comparison. (Any object dominated by y is also dominated by x.)



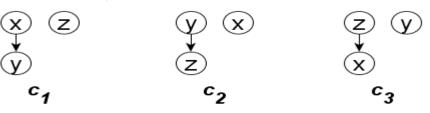
#### Transitivity of Object Dominance: Doesn't Hold

Fundamental reason: lack of explicit attribute representation

- In skyline/preference queries: on any attribute,  $x \ge y$ ,  $y \ge z \implies x \ge z$ .
- o In Pareto-optimal object finding:

 $x \succ_{c} or \sim_{c} y, y \succ_{c} or \sim_{c} z \Longrightarrow x \succ_{c} or \sim_{c} z \text{ (not true)}$ 

Even possible that an object is dominated by only one non-Pareto optimal object.





#### Can Still Benefit From A Similar Idea

For a non-Pareto optimal object, we only need to know at least one object dominates it. We don't care about which other objects also dominate it.

#### Overriding principle of the framework:

- o Identify non-Pareto objects as early as possible
- o Postpone their comparisons with other objects as much as possible



# Candidate Questions

Given asked questions  $Q = \langle q1, ..., qn \rangle$ , x ?<sub>c</sub> y is a candidate question iff it satisfies 3 conditions:

(i) The outcome of  $x ?_c y$  is unknown yet, i.e.,  $rlt(x ?_c y) \notin R^+(Q)$ (ii)  $x \in O$ ? (iii) Based on R+(Q),  $y \succ x$  is not ruled out yet.

How to rule out  $y \succ x$ ?  $\exists c \in C$  such that  $x \succ_c y \in R^+(Q) \Rightarrow y \nvDash x$ 



Only Choosing from Candidate Questions

Sufficient Property 2:  $Qcan = \emptyset \iff O? = \emptyset$ 

#### Efficient

Theorem 1: If Q contains non-candidate questions, there exists a shorter or equally long sequence Q' without non-candidate questions such that Q' finds at least all dominated objects found by Q.



# Macro-ordering, Micro-ordering

Both guided by the overriding principle

Macro-ordering: When available, we choose a candidate question  $x ?_c y$  such that  $y \notin O_x$ Micro-ordering: Several question ordering heuristics:

- Random Question (RandomQ): randomly choose a candidate question  $x ?_c y$
- Random Pair (RandomP): randomly choose a candidate question x ?<sub>c</sub> y, continue to finish all remaining candidate questions between x and y.
- Fewest Remaining Questions (FRQ): Choose a pair with the fewest remaining questions. Ties are broken based on how many objects are better/worse than x and y on the criterion.



# Experiments by Simulation

• Used an 10000-tuple NBA dataset that records players' perseason performance on 10 criteria (points-per-game, ...)

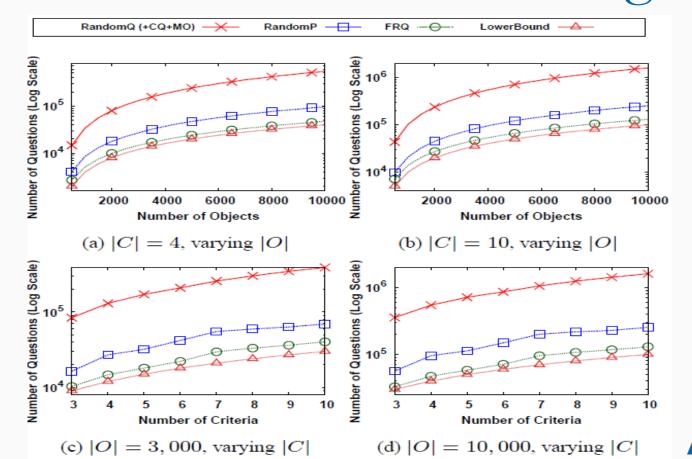
• Simulated Partial orders based on players' performance comparison, with some perturbations.



#### Effectiveness of Candidate Questions and Macro-Ordering

Number of Questions (Log Scale)  $_{\phi}$  01  $_{0}$  01  $_{0}$  02  $_{0}$  01  $_{$ 2000 4000 6000 8000 10000 500 2000 2500 3000 1000 1500 Number of Objects Number of Objects (a) |C| = 4, varying |O|(b) |C| = 10, varying |O|Number of Questions (Log Scale) 10<sup>8</sup> 107 10<sup>6</sup> 10<sup>6</sup> 105 9 10 5 8 9 10 3 8 Number of Criteria Number of Criteria (c) |O| = 3,000, varying |C|(d) |O| = 10,000, varying |C|

## Effectiveness of Micro-Ordering





### Experiments using Amazon Mechanical Turk

- Compare 100 photos of UT-Arlington campus, by color, sharpness, landscape.
- All 14, 850 possible pairwise questions were partitioned into 1, 650 tasks, each containing 9 questions on a criterion.
- Worker qualification:
- responded to at least 100 HITs before with at least 90% approval rate
- 2 additional validation questions mixed in each task



#### Experiments using Amazon Mechanical Turk

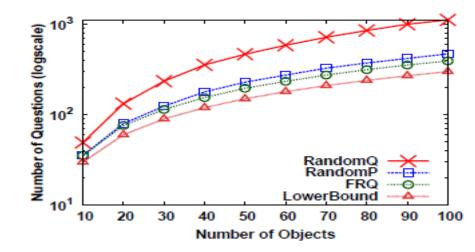


Figure 8: No. of questions by different micro-ordering heuristics. |C| = 3, varying |O|.



### Limitations and Future Work

No performance guarantee: as bad as brute-force in worst-case Non-deterministic results: due to contradiction resolution Possibly empty result: due to lack of object dominance transitivity No consideration of different levels of confidence on question outcomes or crowdsourcers' quality. Future work: Pareto-optimal objects in probabilistic sense? No consideration of parallel/batch-execution scheme Future work: Parallel scheme



## Thank You! Questions?

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(3) Contradiction Resolution How often does contradiction happen?

- o Depends on data itself, k, and  $\theta$
- o It may not happen a lot.
  - □ Intuitively: as long as the underlying relation is transitive, collective wisdom of crowd should reflect it.
  - Preference judgments of relevance in document retrieval are transitive [27, 11].



Brute-Force on the Toy Example

### 6 objects, 3 criteria: 45 comparisons



#### RandomQ on the Toy Example

i	$rlt(q_i)$	derived results	$O_{}$	$O_?$	$O_{ imes}$
1-9	b≻ <u>m</u> e, <u>c</u> ∼ <sub>a</sub> d, <u>a</u> ∼ <sub>m</sub> c		Ø	{a,b,c,d,e,f}	Ø
	$\underline{\mathbf{c}}\succ_s \mathbf{e}, \underline{\mathbf{b}}\sim_s \mathbf{d}, \mathbf{b}\succ_a \underline{\mathbf{a}}$				
	$\underline{d} \succ_m e, \underline{b} \sim_m d, \underline{b} \succ_s f$				
10	<u>a</u> ≻ <sub>m</sub> d	a≻me			
11	<b>c</b> ≻_a <u>a</u>				
12	$\underline{b}{\sim_s}c$				
13	c≻ <u>m</u> <u>d</u>	$c\succ_me$			
14 - 19	d≻ <u>₅e</u> , <u>e</u> ~ <sub>a</sub> c, <u>d</u> ~ <sub>a</sub> f				
	<u>a</u> ∼ <sub>a</sub> d, f≻ <sub>a</sub> <u>a</u> , <u>b</u> ≻ <sub>a</sub> e				
20	b≻ <u>a</u> d	p≻q	Ø	{a,b,c,e,f}	{d}
21 - 23	c≻ <u>s</u> <u>f</u> , a≻ <u>s</u> <u>e</u> , <u>f</u> ∼mb				
24	a≻ <u>₅f</u>	a∼f			
25	<u>e</u> ≻af	b≻af, b≻aa	Ø	{a,b,c,e}	{d,f}
		e≻ <sub>a</sub> a, b≻f			
26	<u>b</u> ≻ac				
27	a≻ <u>mb</u>				
28	b≻ <u>s</u> e_	b≻e	Ø	{a,b,c}	{d,e,f}
29	<u>c</u> ≻ <sub>s</sub> a	c≻₅e, c≻a	Ø	{b,c}	{a,d,e,f}
30	<u>b</u> ∼mc	p≻c	{b}	Ø	{a,c,d,e,f}



#### RandomP on the Toy Example

i	$rlt(q_i)$	derived results			$O_{ imes}$
1	c≻ <u>₅f</u>		Ø	{a,b,c,d,e,f}	Ø
2	<u>f</u> ≻mc	f∼c			
	<u>a</u> ≻ <sub>s</sub> e, <u>a</u> ≻ <sub>m</sub> e				
5	e≻ <u>a</u>	$a{\sim}e$			
6 - 7	c≻ <u>₅e</u> , c≻ <u>m</u> e				
8	$\underline{e}{\sim_a}c$	c≻e	Ø	{a,b,c,d,f}	{e}
9	<u>b</u> ≻ <sub>s</sub> a	$b\succ_se$			
10	a≻ <u>mb</u>	a∼b			
11	<u>d</u> ≻ <sub>s</sub> f				
12	f≻ <u>m</u> d	f∼d			
13	d≻ <u>sa</u>	$d_{\succ_s}e$			
14	<u>a</u> ≻ <sub>m</sub> d	a∼d			
	$\underline{b} \sim_s c, \underline{b} \sim_m c$				
17	<u>b</u> ≻ <sub>a</sub> c	p≻c	Ø	{a,b,d,f}	{c,e}
	$\underline{d} \sim_s b, \underline{d} \sim_m b$				
20	b≻ <u>a</u> d	p≻q	Ø	{a,b,f}	{ <b>c</b> ,d,e}
21	<u>a</u> ≻₅f	$b \succ_s f$			
22	$\underline{a} \sim_m f$				
23	f≻ <u>a</u>	a∼f			
24	b∼ <u>m</u> f				
25	b≻ <u>a</u> f	b≻ <sub>a</sub> a, b≻f	{b}	{a}	{c,d,e,f}
	c≻ <u>₅a</u> , <u>a</u> ~mc				
28	c≻ <u>a</u> a	c≻a	{b}	Ø	{a,c,d,e,f}



#### FRQ on the Toy Example

i	$rlt(q_i)$	derived results	(x,y), C <sub>X,y</sub>	$O_{}$	$O_?$	$O_{ imes}$
			$(a,b), \{s,m,a\}$	Ø	{a,b,c,d,e,f}	Ø
1	b≻ <u>₅a</u>		$(a,b), \{m,a\}$			
<b>2</b>	<u>a</u> ≻ <sub>m</sub> b	a∼b	(a,c), $\{s,a,m\}$			
3	c≻ <u>₅a</u>		$(a,c), \{a,m\}$			
4	<u></u>		$(a,c), \{m\}$			
	c≻ <u>m</u> a	c≻a	(b,c), $\{a,s,m\}$	Ø	$\{b,c,d,e,f\}$	{a}
	$\underline{b} \sim_a \mathbf{c}$		(b,c), $\{s,m\}$			
7	$\underline{b}\sim_s c$		$(b,c), \{m\}$			
		b≻ <i>m</i> a, b≻c	(d,b), $\{a,s,m\}$	Ø	{b,d,e,f}	{a,c}
9	<u> </u>		$(d,b), \{s,m\}$			
	$b \sim_s \underline{d}$		$(d,b), \{m\}$			
	b≻ <u>m</u> d	p≻q	(e,b), $\{a,s,m\}$	Ø	{b,e,f}	{a,c,d}
	b≻ <u>a</u> e		$(e,b), \{s,m\}$			
13	b≻ <u>₅e</u>		(e,b), $\{m\}$			
		a≻ <i>m</i> e, b≻e	(f,b), $\{a,s,m\}$	Ø	{b, <b>f</b> }	a,c,d,e
	b≻ <u>a</u> f		(f,b), $\{s,m\}$			
	b≻ <u>sf</u>		(f,b), $\{m\}$			
17	b≻ <u>m</u> f	b≻f		{b}	Ø	{a,c,d,e,f}

