



# **Multi-Criteria Decision Support using ELECTRE methods**

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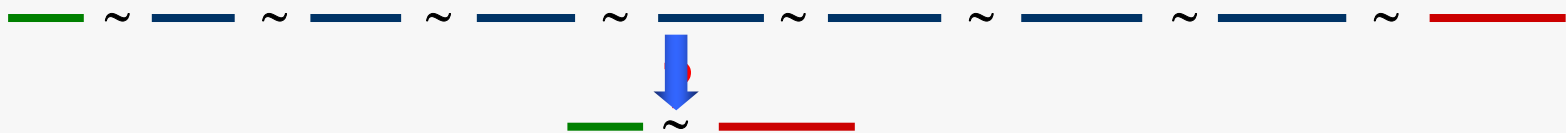
# Weak points of preference modelling using utility (value) function

- Utility function distinguishes only 2 possible relations between actions:

preference relation:  $a \succ b \Leftrightarrow U(a) > U(b)$

indifference relation:  $a \sim b \Leftrightarrow U(a) = U(b)$

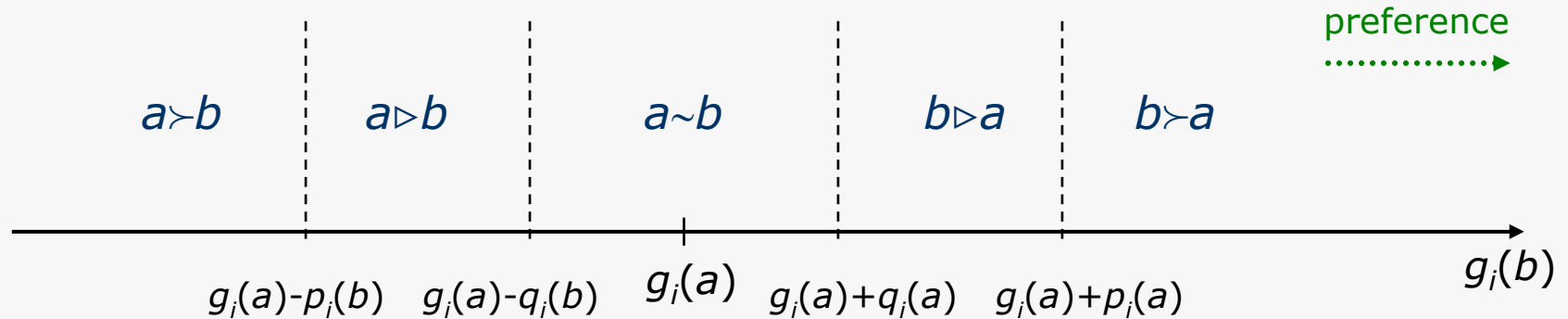
- $\succ$  is asymmetric (antisymmetric and irreflexive) and transitive
- $\sim$  is symmetric, reflexive and transitive
- Transitivity of indifference is troublesome, e.g.



- In consequence, a non-zero indifference threshold  $q_i$  is necessary
- An immediate transition from indifference to preference is unrealistic, so a preference threshold  $p_i \geq q_i$  and a weak preference relation  $\triangleright$  are desirable
- Another realistic situation which is not modelled by  $U$  is incomparability, so a good model should include also an incomparability relation ?

# Four basic preference relations and an outranking relation $S$

- Four basic preference relations are:  $\{\sim, \triangleright, \succ, ?\}$



- Criterion with thresholds  $p_i(a) \geq q_i(a) \geq 0$  is called **pseudo-criterion**
- The four basic situations of indifference, strict preference, weak preference, and incomparability are sufficient for establishing a realistic model of Decision Maker's (DM's) preferences

# Four basic preference relations and an outranking relation $S$

- **Axiom of limited comparability** (Roy 1985):

*Whatever the actions considered, the criteria used to compare them, and the information available, one can develop a satisfactory model of DM's preferences by assigning one, or a grouping of two or three of the four basic situations, to any pair of actions.*

- **Outranking relation**  $S$  groups three basic preference relations:

$$S = \{\sim, \triangleright, \succ\} \text{ – reflexive and non-transitive}$$

$aSb$  means: „**action  $a$  is at least as good as action  $b$** ”

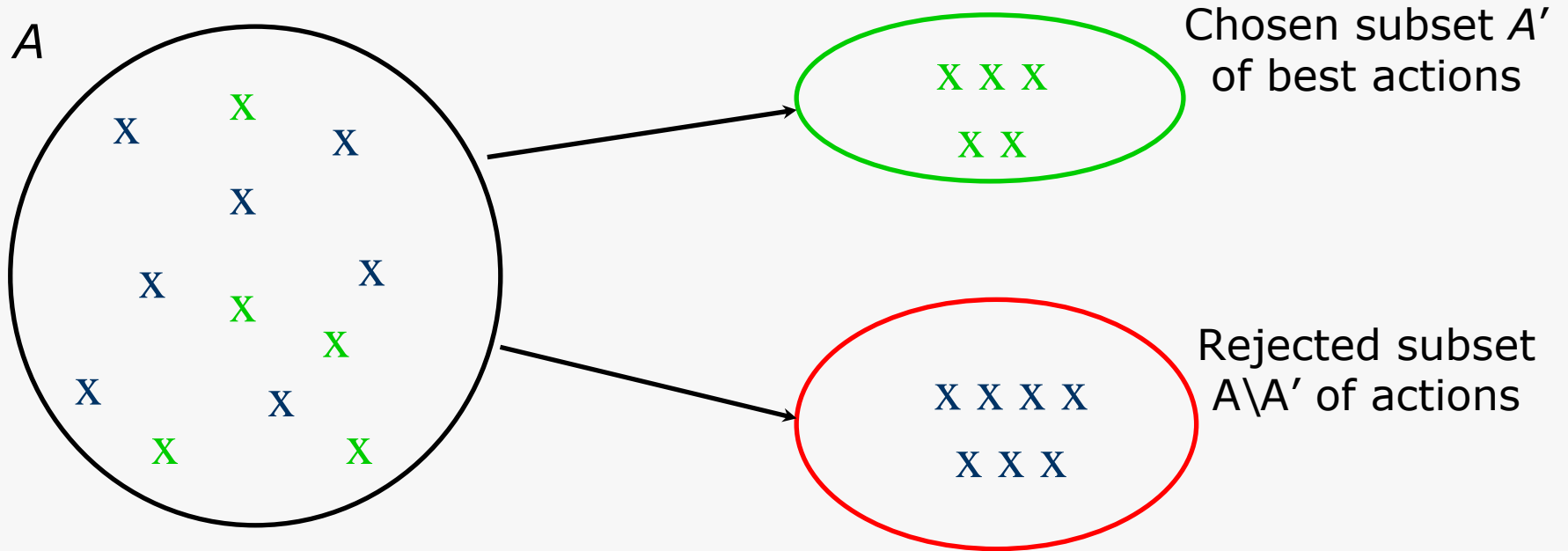
- For each couple  $a, b \in A$ :

$$aSb \wedge \text{non } bSa \Leftrightarrow a \triangleright b \vee a \succ b$$

$$aSb \wedge bSa \Leftrightarrow a \sim b$$

$$\text{non } aSb \wedge \text{non } bSa \Leftrightarrow a ? b$$

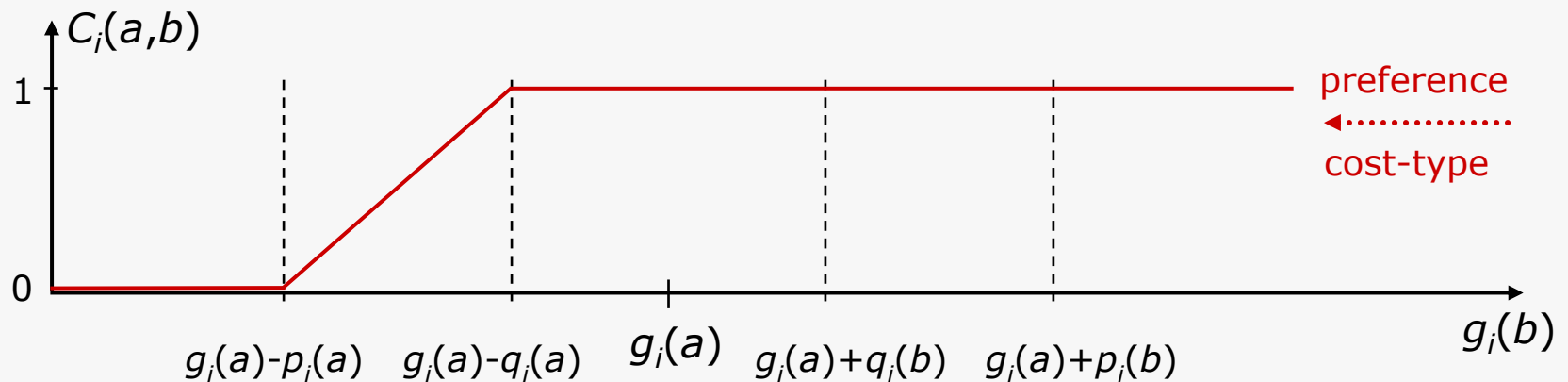
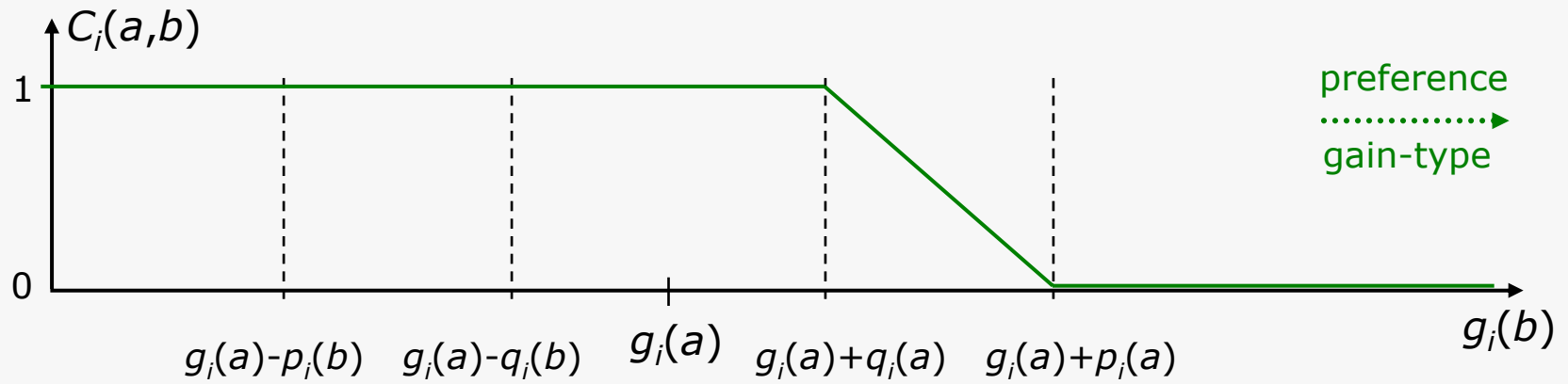
## ELECTRE Is: choice problem ( $P_\alpha$ )



- **Input data**: finite set of actions  $A = \{a, b, c, \dots, h\}$   
consistent family of criteria  $G = \{g_1, g_2, \dots, g_n\}$
- **Preferential information** ( $i=1, \dots, k$ )
  - intracriteria: – indifference thresholds  $q_i(a) = \alpha_i^q \times g_i(a) + \beta_i^q$   
– preference thresholds  $p_i(a) = \alpha_i^p \times g_i(a) + \beta_i^p$
  - intercriteria: – importance coefficients (weights) of criteria  $k_i$   
– veto thresholds  $v_i(a) = \alpha_i^v \times g_i(a) + \beta_i^v$
- $0 \leq q_i(a) \leq p_i(a) \leq v_i(a)$  are functions of a **worse** evaluation of the two being compared
- The **preference model**, i.e. outranking relation  $S$  in set  $A$  is constructed via **concordance** and **discordance tests**

# ELECTRE Is - concordance test

- Checks if the coalition of criteria concordant with the hypothesis  $aSb$  is strong enough (for each couple  $a, b \in A$ )
- Concordance coefficient  $C_i(a, b)$  for each criterion  $g_i$ :



## ELECTRE Is - concordance test

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- Aggregation of concordance coefficients for  $a, b \in A$ :

$$C(a, b) = \frac{\sum_{i=1}^n k_i C_i(a, b)}{\sum_{i=1}^n k_i}$$

- $C(a, b) \in [0, 1]$
- Concordance test is positif if:  $C(a, b) \geq \lambda$

where  $\lambda$  is a cutting level, such that  $0.5 \leq \lambda \leq 1 - \min_{i=1, \dots, n} \{k_i\} / \sum_{i=1}^n k_i$

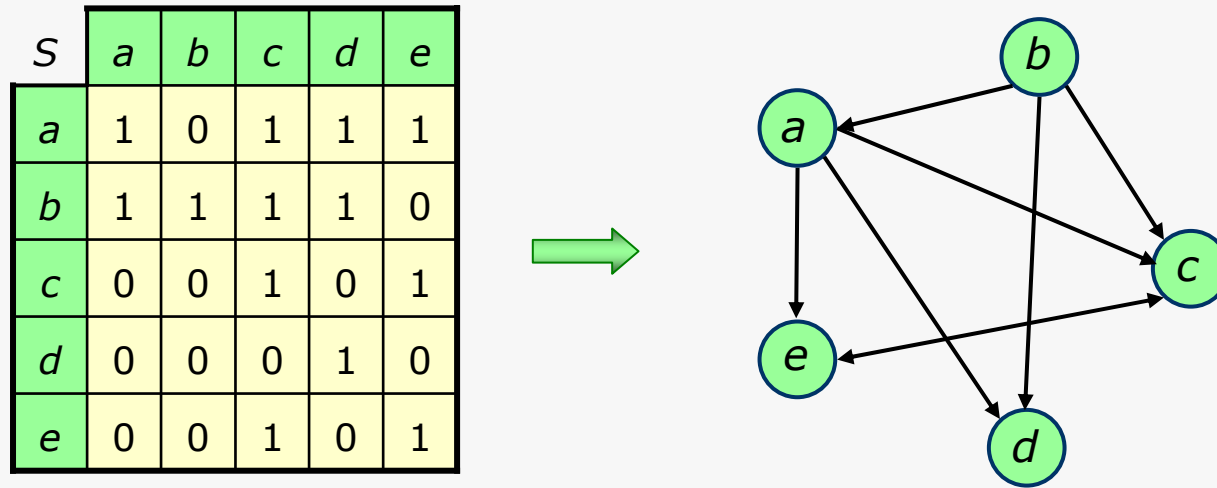


## ELECTRE Is - discordance test and outranking relation

- Checks if among criteria discordant with the hypothesis  $aSb$  there is a strong opposition against  $aSb$  (for each  $a, b \in A$  such that  $C(a, b) \geq \lambda$ )
- For a discordant criterion  $g_i$ , the **opposition is strong** if:
  - (for **gain-type** criterion)  $g_i(b) - g_i(a) \geq v_i(a)$
  - (for **cost-type** criterion)  $g_i(a) - g_i(b) \geq v_i(a)$
- **Conclusion**:  $aSb$  is true if and only if  $C(a, b) \geq \lambda$  and there is no criterion strongly opposed to the hypothesis
- In result, for each couple  $a, b \in A$ , one obtains relation  $S$  either **true** (1) or **false** (0)
- The **preference structure in set A** can be represented by a **graph** where nodes represent actions and arcs represent relation  $S$

# ELECTRE Is – exploitation of the outranking graph

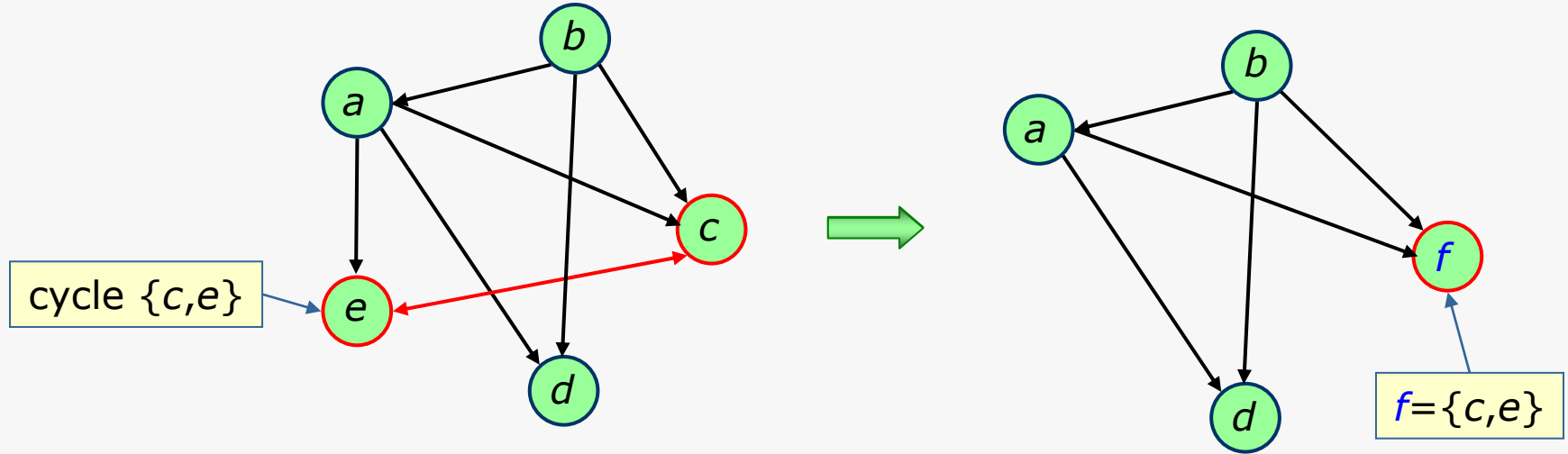
- The **outranking graph**  $S$  represents a preference structure in  $A$



- What are **the best actions** ?
- **Kernel**  $K$  of the outranking graph:
  - actions (nodes) belonging to  $K$  **do not outrank each other** (no arc between them)
  - each action not belonging to  $K$  **is directly outranked** by at least one action from  $K$

## ELECTRE Is – search for the kernel

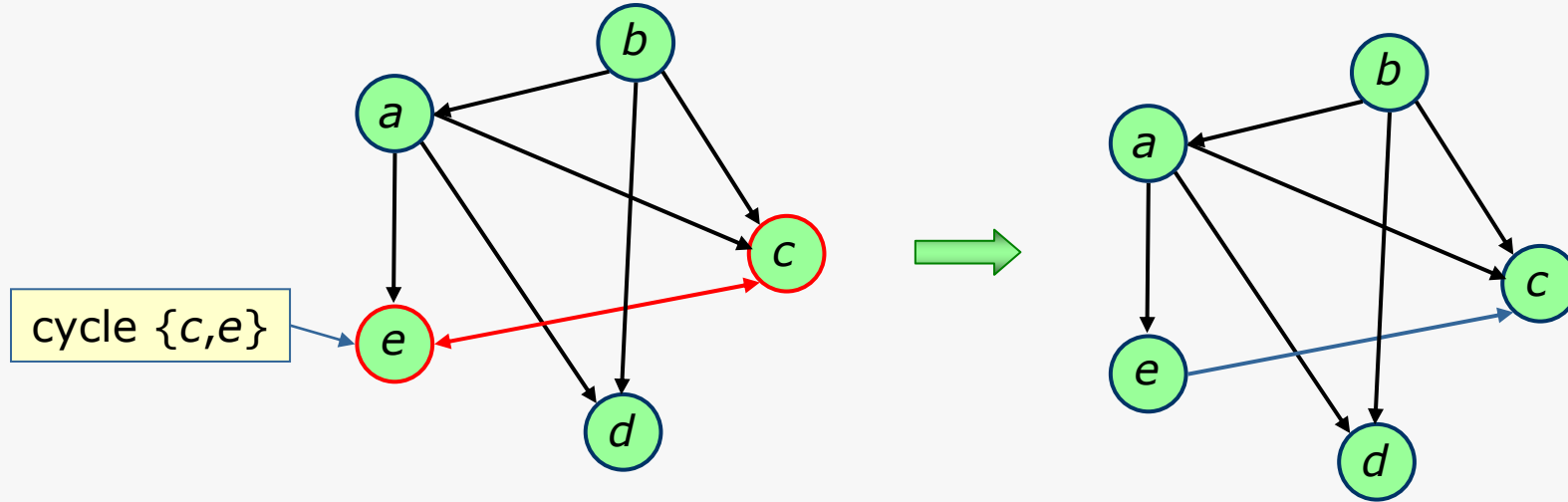
- In order to have a **single kernel**, the graph should be acyclic



- **Cycles** should be first eliminated in one of two ways:
  - replacing the cycle by a **dummy node** representing an equivalence class (clique)

# ELECTRE Is – search for the kernel

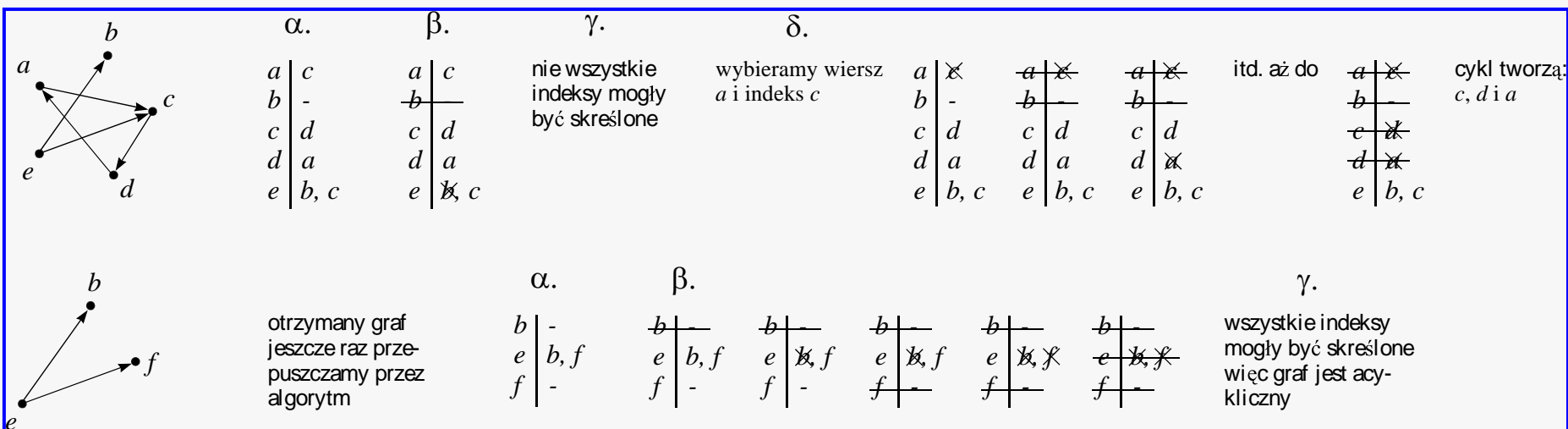
- In order to have a **single kernel**, the graph should be acyclic



- **Cycles** should be first eliminated in one of two ways:
  - replacing the cycle by a **dummy node** representing an equivalence class (clique)
  - **cutting** the cycle = eliminating the „weakest” arcs (outranking relations)

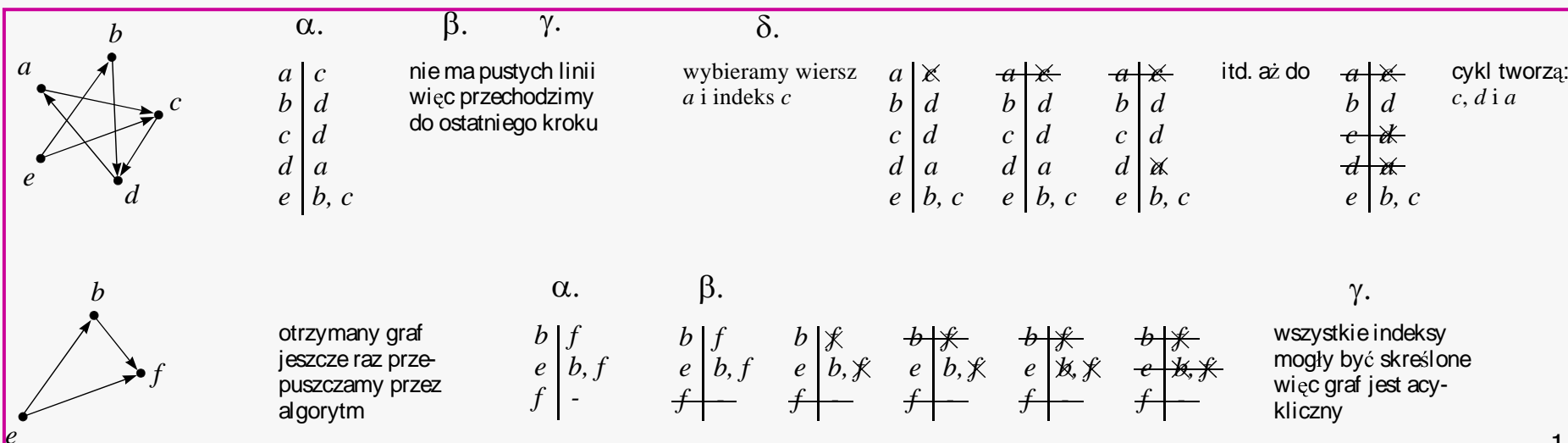
# Algorytm wykrywania cykli w grafie

- α. Utworzyć tablicę z jednym wejściem. W każdym wierszu wpisać indeksy bezpośrednich **następników** danego wierzchołka  $x_i$  (tzw. **listę ZA**).
- β. Poszukać linii pustą i skreślić numer tej linii wszędzie tam gdzie występuje w tablicy. Linie z samymi skreślonymi indeksami traktować jak puste. Skreślanie kontynuować aż do wyczerpania możliwości.
- γ. Jeśli wszystkie indeksy mogły być skreślone, to graf nie ma cykli. W przeciwnym razie ma co najmniej jedną cykl.
- δ. Wybrać indeks dowolnej linii niepustej a w niej dowolny nie skreślony indeks i skreślić go. Skreślanie kontynuować do momentu, gdy sekwencja skreślonych indeksów utworzy **cykl**.



# Algorytm wykrywania cykli w grafie

- α. Utworzyć tablicę z jednym wejściem. W każdym wierszu wpisać indeksy bezpośrednich **następników** danego wierzchołka  $x_i$  (tzw. **listę ZA**).
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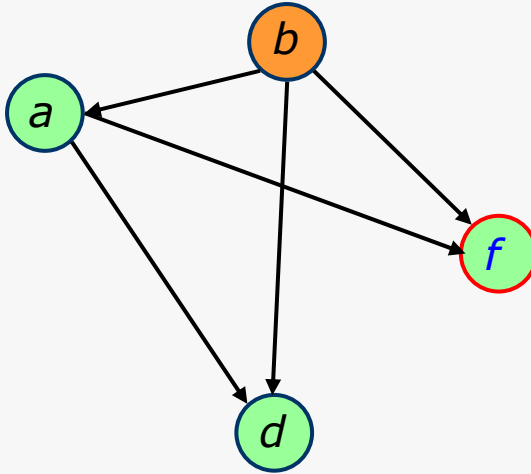
# Algorytm znajdowania jądra w grafie acyklicznym

- α. Utworzyć tablicę z jednym wejściem. W każdym wierszu wpisać indeksy bezpośrednich **poprzedników** danego wierzchołka  $x_i$  (tzw. **listę PRZED**).
- β. Zaznaczyć linię pustą (znaczką  $\times$ ). Skreślić całe linie, w których występuje wierzchołek z indeksem zaznaczonej linii pustej.  
 Skreślić indeks skreślonego wiersza gdziekolwiek pojawia się w tablicy.  
 Każda linia z samymi skreślonymi indeksami jest uznawana za pustą.  
 Iterować aż do wyczerpania możliwości.
- γ. Jeśli wszystkie indeksy mogły być zaznaczone lub skreślone, to graf ma **pojedyncze jądro** złożone z wierzchołków odpowiadających zaznaczonym ( $\times$ ) wierszom.

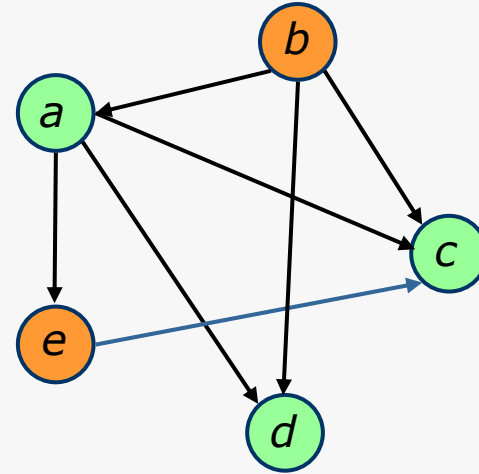
	α.	β.					γ.
	$\begin{array}{c c} a & - \\ b & d \\ c & a, d, e \\ d & a \\ e & b \end{array}$	$\begin{array}{c c} \times a & - \\ b & d \\ c & a, d, e \\ d & a \\ e & b \end{array}$	$\begin{array}{c c} \times a & - \\ b & d \\ \cancel{c} & \cancel{a, d, e} \\ \cancel{d} & \cancel{a} \\ e & b \end{array}$	$\begin{array}{c c} \times a & - \\ b & \cancel{d} \\ \cancel{c} & \cancel{a, d, e} \\ \cancel{d} & \cancel{a} \\ e & b \end{array}$	$\begin{array}{c c} \times a & - \\ \times b & \cancel{d} \\ \cancel{c} & \cancel{a, d, e} \\ \cancel{d} & \cancel{a} \\ e & b \end{array}$	$\begin{array}{c c} \times a & - \\ \times b & \cancel{d} \\ \cancel{c} & \cancel{a, d, e} \\ \cancel{d} & \cancel{a} \\ \cancel{e} & \cancel{b} \end{array}$	wszystkie wiersze mogły zostać zaznaczone lub skreślone, więc graf ma pojedyncze jądro: $N = \{a, b\}$

# ELECTRE Is – final recommendation

- The best actions = the kernel:



$$K=\{b\}$$



$$K=\{b,e\}$$

- Actions in the kernel can be:
  - **initial** (without predecessors)
  - **isolated** (incomparable to all others)
  - **intermediate** (both outranking and outranked)
  - **final** (without successors)



# ELECTRE Is - example

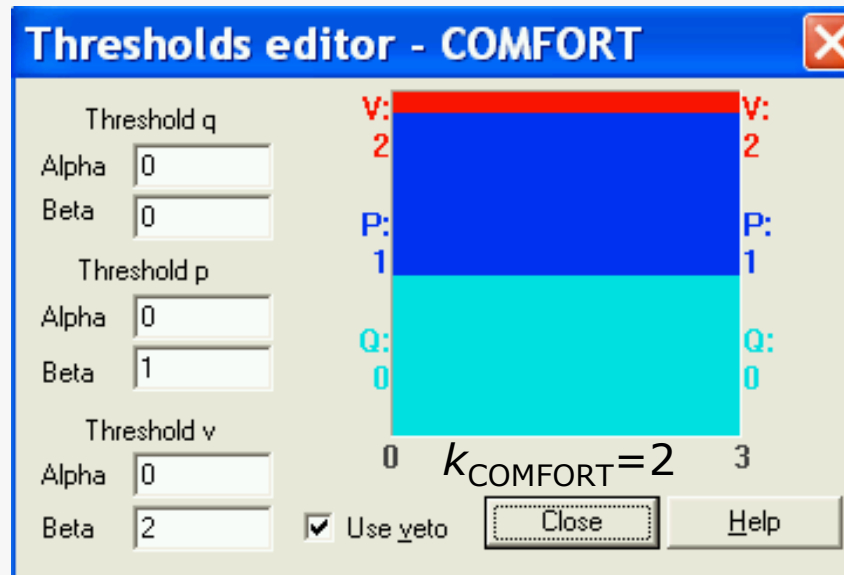
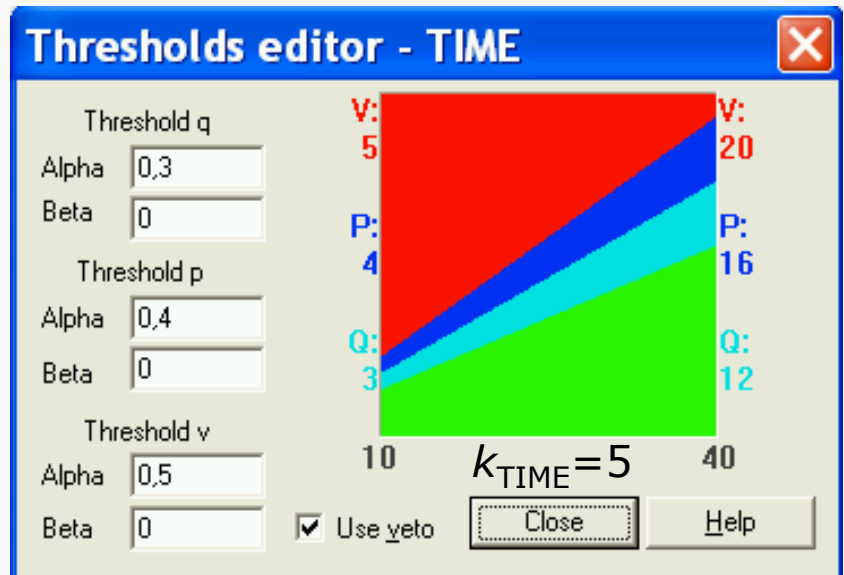
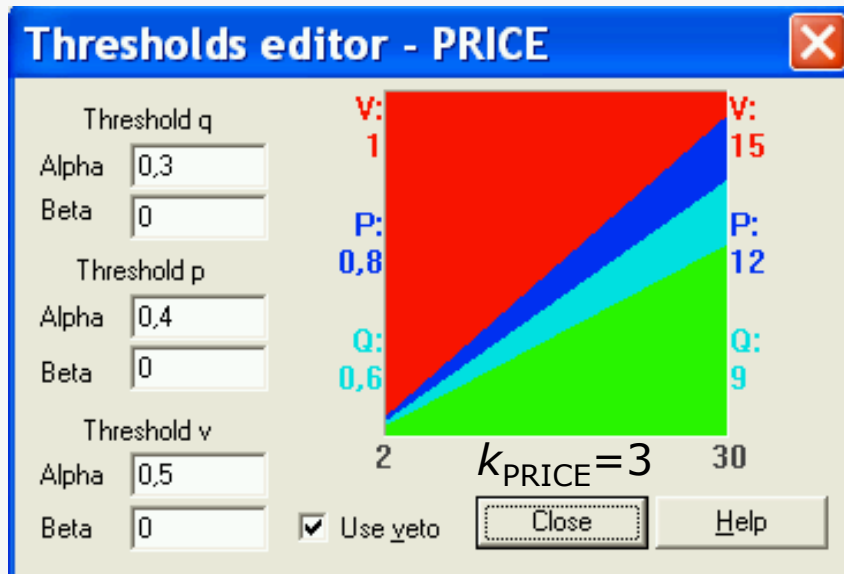
**Electre 1s 2002 - Current project**

File View Alternatives Attributes Electre1s Options Window

**"Metro" Metro.xml**

Alternatives Concordance Discordance Robustness Outranking

	PRICE	TIME	COMFORT	
RER		3	10	1
METRO_1	4	20		2
METRO_2	2	20		0
BUS	6	40		0
TAXI	30	30		3
SNCF	3	20		2



# ELECTRE Is - example

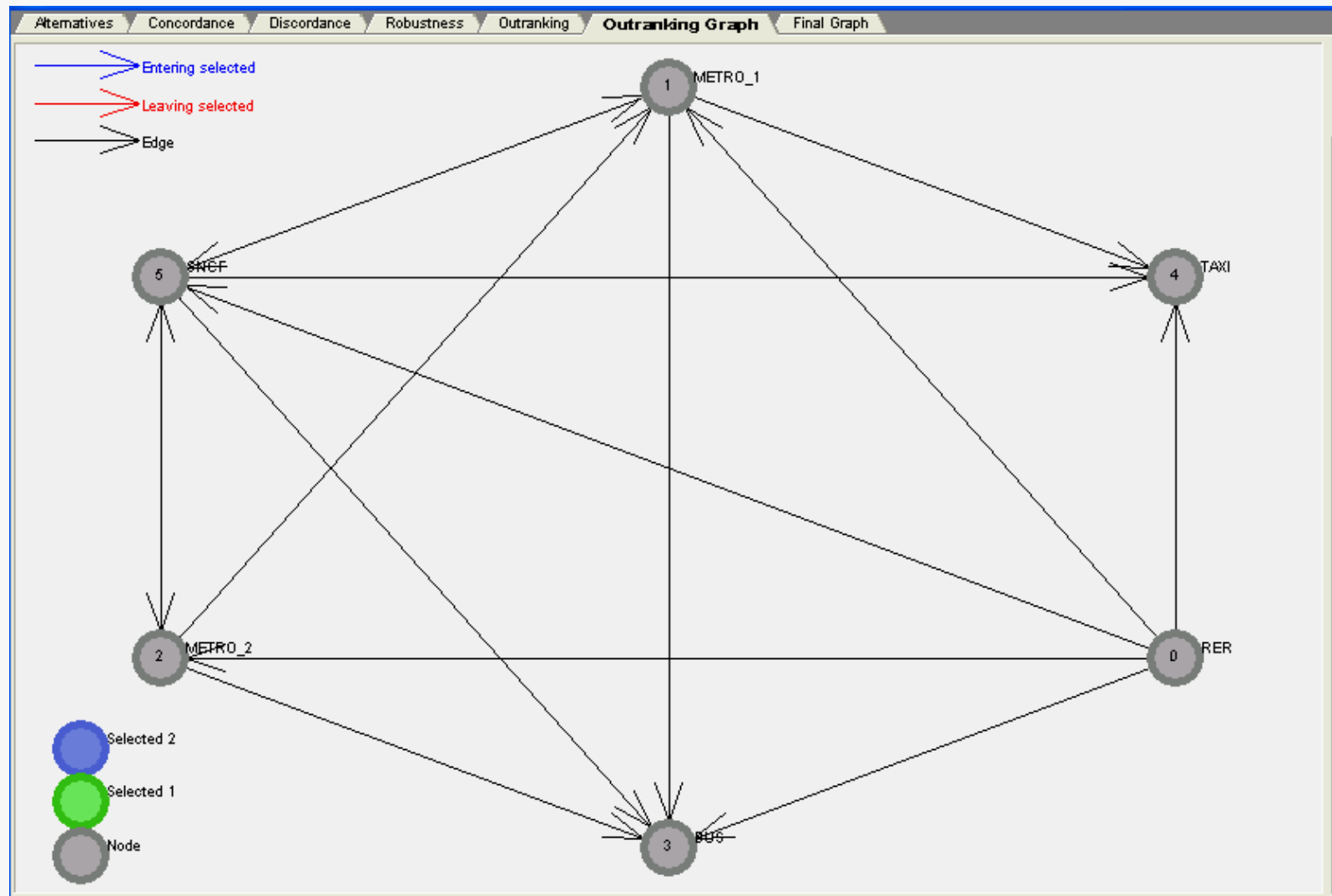
Alternatives	Concordance	Discordance	Robustness	Outranking	Outranking Graph	Final Graph
	RER	METRO_1	METRO_2	BUS	TAXI	SNCF
RER	1	0,8	0,9	1	0,8	0,8
METRO_1	0,5	1	0,7	1	0,8	1
METRO_2	0,3	0,8	1	1	0,8	0,8
BUS	0	0,2	0,2	1	0,8	0
TAXI	0,2	0,533333333333	0,533333333333	0,7	1	0,533333333333
SNCF	0,5	1	0,9	1	0,8	1

$$\lambda=0.75$$

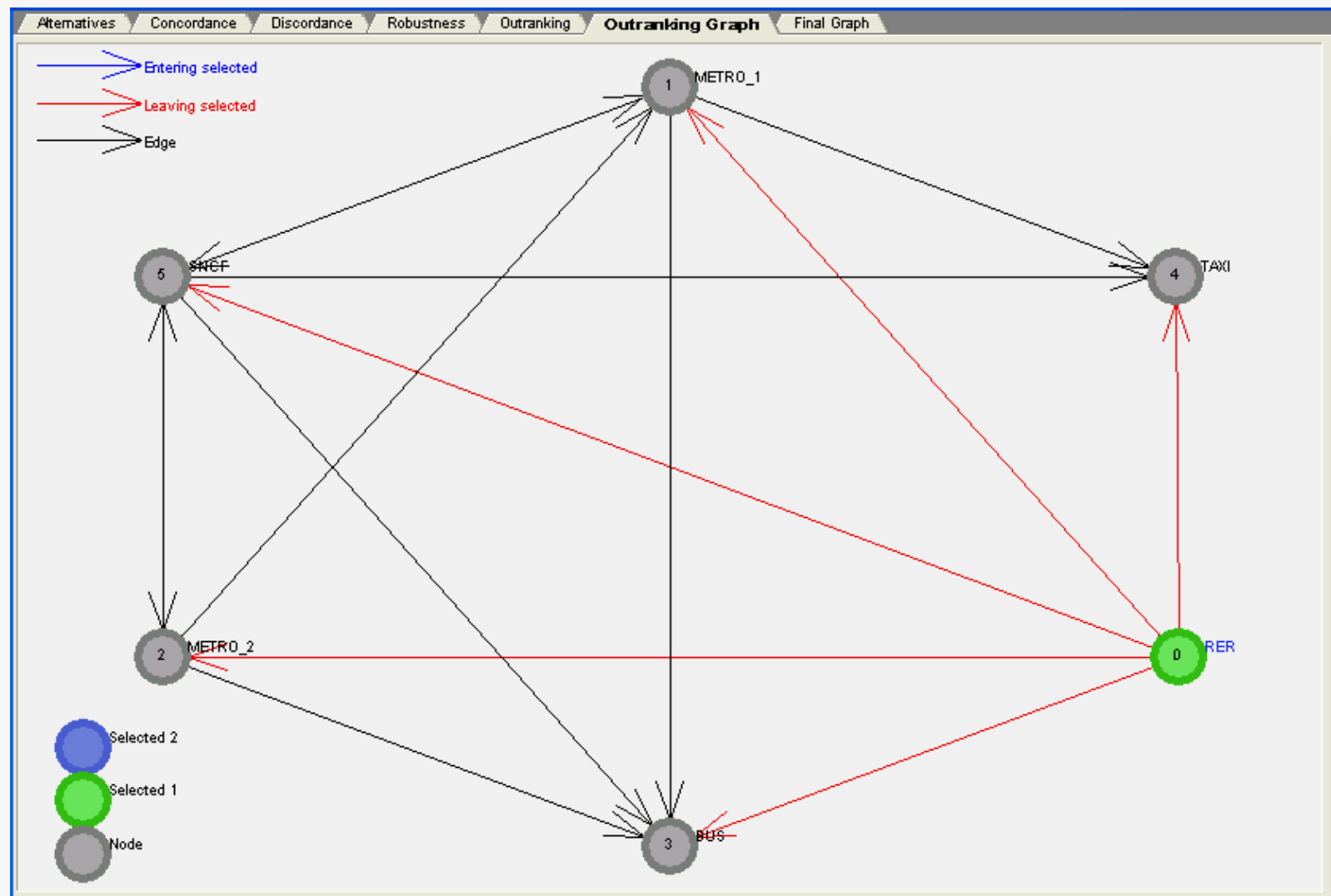
Alternatives	Concordance	Discordance	Robustness	Outranking	Outranking Graph	Final Graph
	RER	METRO_1	METRO_2	BUS	TAXI	SNCF
RER	0	0	0	0	0	0
METRO_1	-	0	-	0	0	0
METRO_2	-	0	0	0	1	0
BUS	-	-	-	0	1	-
TAXI	-	-	-	-	0	-
SNCF	-	0	0	0	0	0

Alternatives	Concordance	Discordance	Robustness	Outranking	Outranking Graph	Final Graph
	RER	METRO_1	METRO_2	BUS	TAXI	SNCF
RER	1	1	1	1	1	1
METRO_1	0	1	0	1	1	1
METRO_2	0	1	1	1	0	1
BUS	0	0	0	1	0	0
TAXI	0	0	0	0	1	0
SNCF	0	1	1	1	1	1

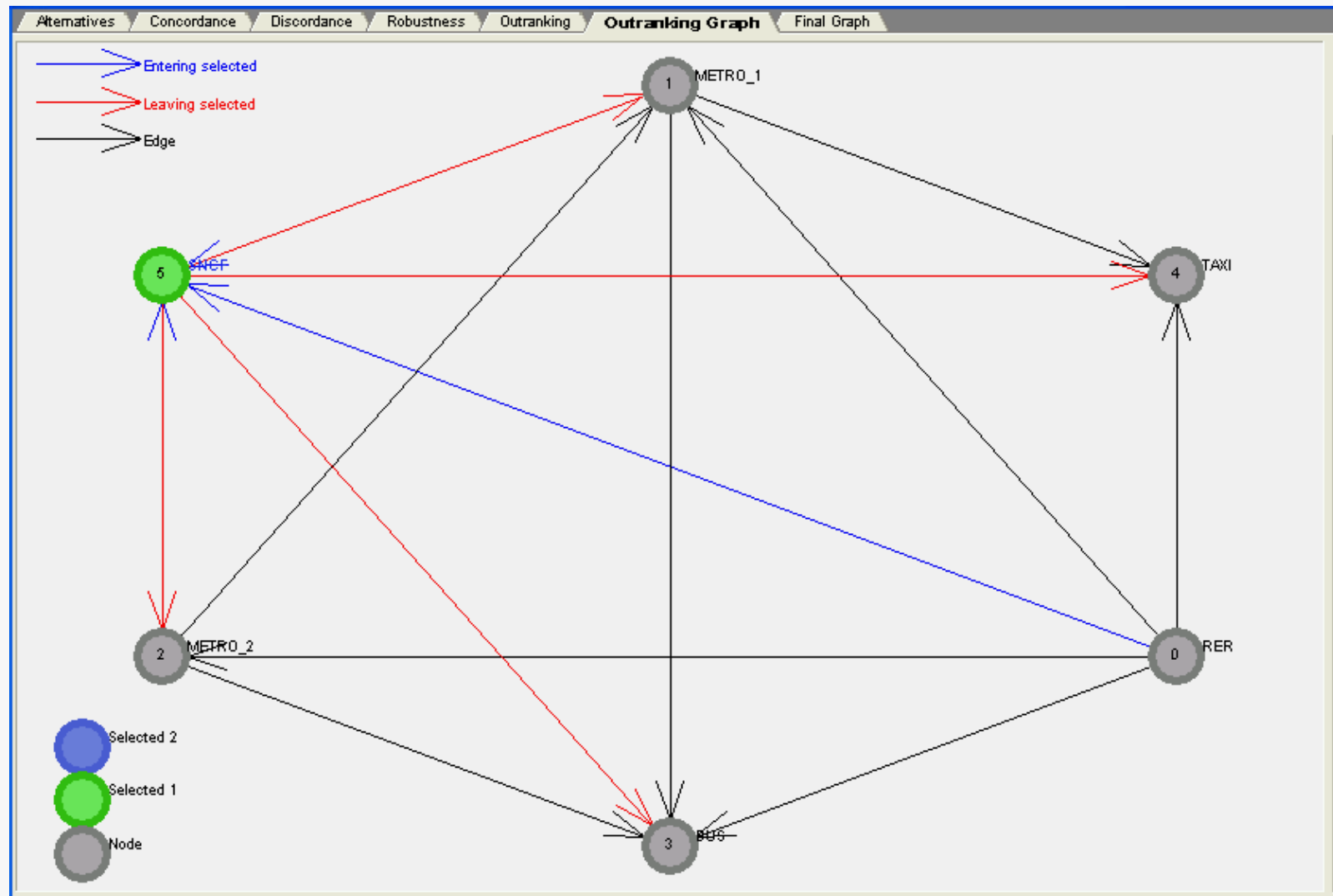
# ELECTRE Is - example



# ELECTRE Is - example



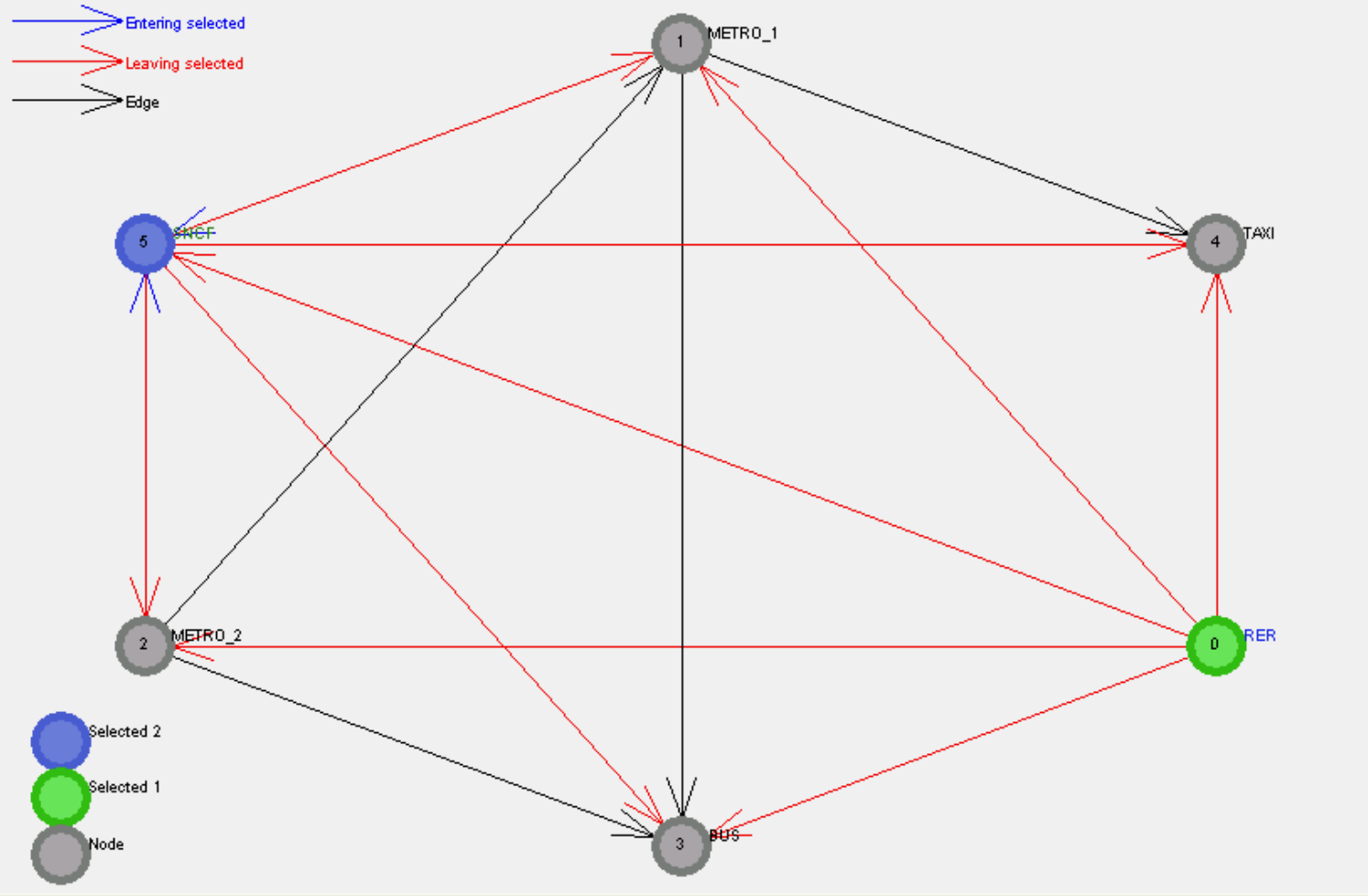
# ELECTRE Is - example



# ELECTRE Is - example

Alternatives   Concordance   Discordance   Robustness   Outranking   **Outranking Graph**   Final Graph

 Entering selected  
 Leaving selected  
 Edge

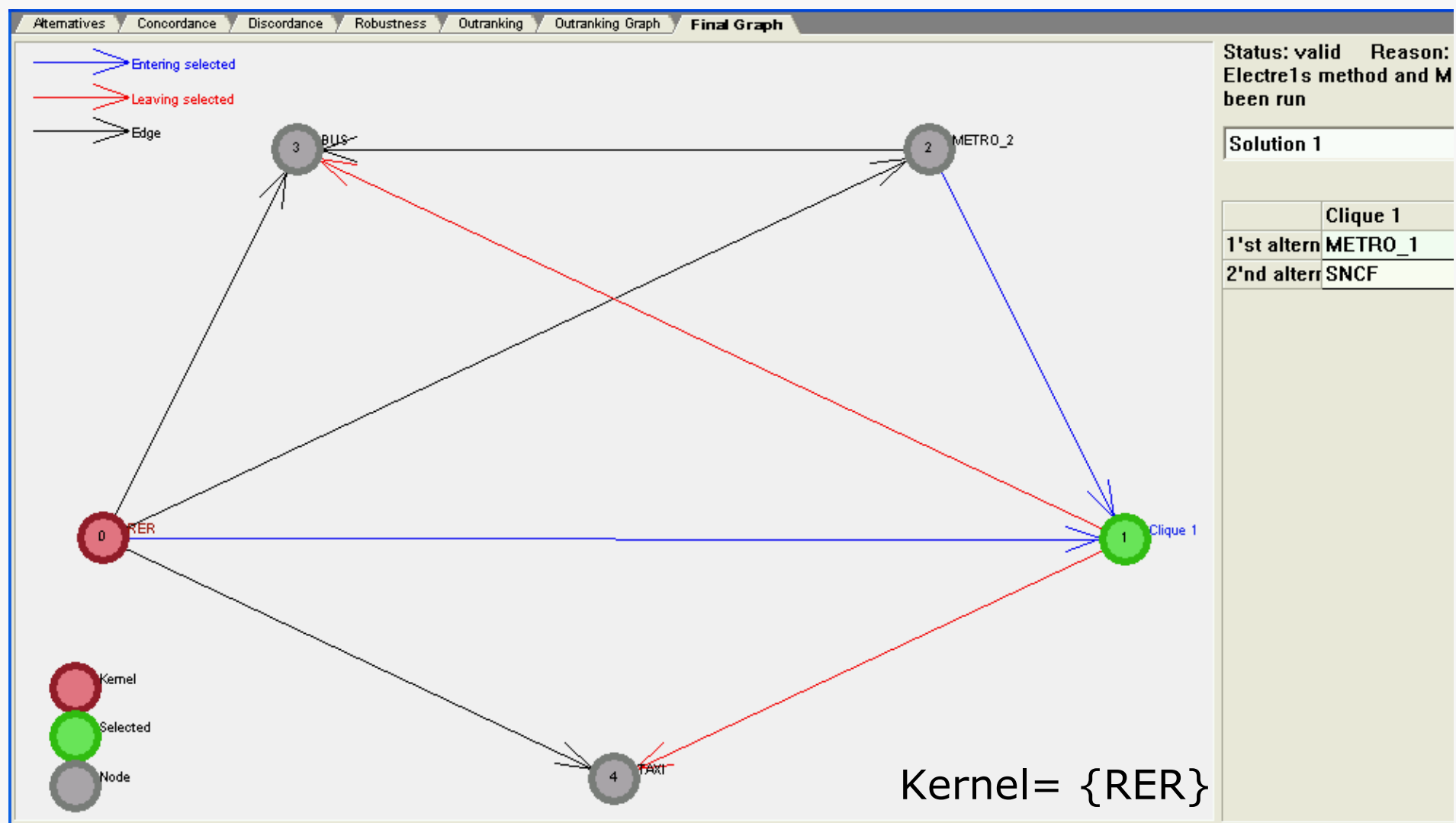


Status: valid   Reason:  
 Electre1s method and MECR have  
 been run

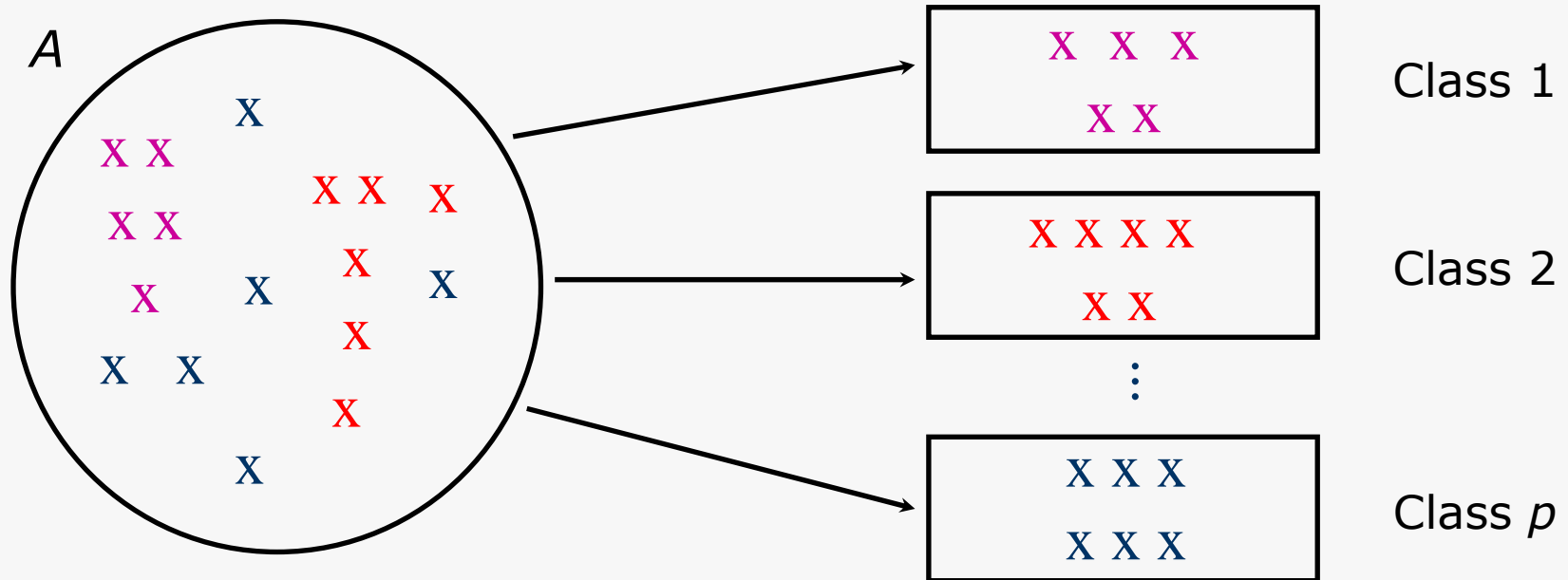
	PRICE	TIME	COMFORT
RER	3	10	1
?	1	1	0
SNCF	3	20	2
Ci[a,b]	1	1	0
Di[a,b]	0	0	0
Ci[b,a]	1	0	1
Di[b,a]	0	0	0

 Selected 2  
 Selected 1  
 Node

# ELECTRE Is - example



# ELECTRE TRI: sorting problem ( $P_\beta$ )

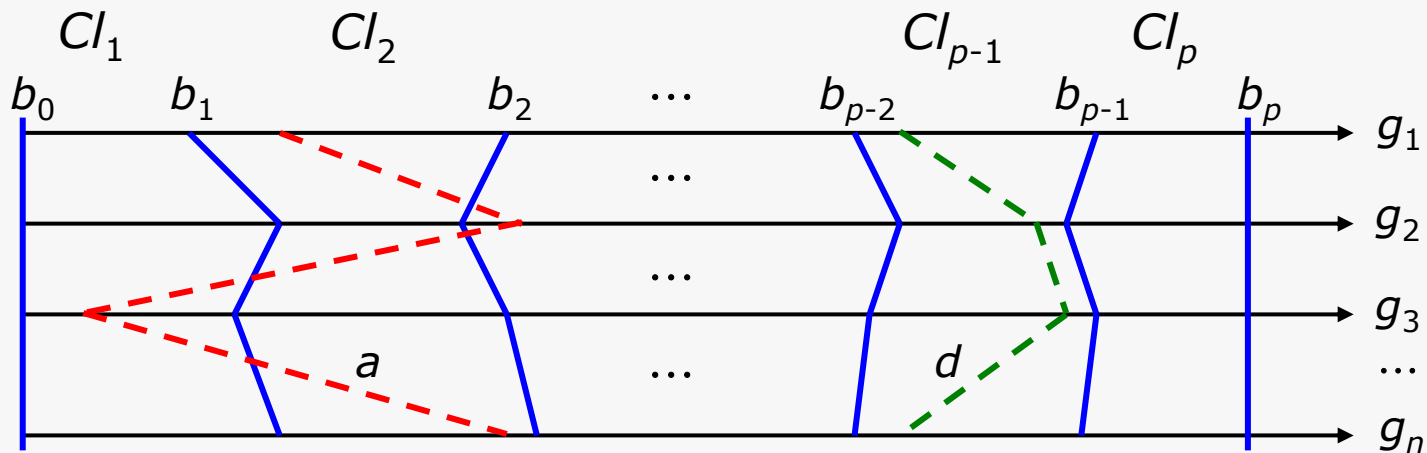


Class 1  $\succ$  Class 2  $\succ$  ...  $\succ$  Class  $p$



- **Input data**: finite set of actions  $A = \{a, b, c, \dots, h\}$   
 consistent family of criteria  $G = \{g_1, g_2, \dots, g_n\}$   
 preference-ordered decision classes  $Cl_t, t=1, \dots, p$

Decision classes are characterized by **limit profiles**  $b_t, t=0, 1, \dots, p$



- The **preference model**, i.e. outranking relation  $S$  is constructed for each couple  $(a, b_t)$ , for every  $a \in A$  and  $t=0, 1, \dots, p$

- **Preferential information** ( $i=1,\dots,k$ )

intracriteria: – indifference thresholds  $q_i^t$  for each class limit,  $t=0,1,\dots,p$   
– preference thresholds  $p_i^t$  for each class limit,  $t=0,1,\dots,p$

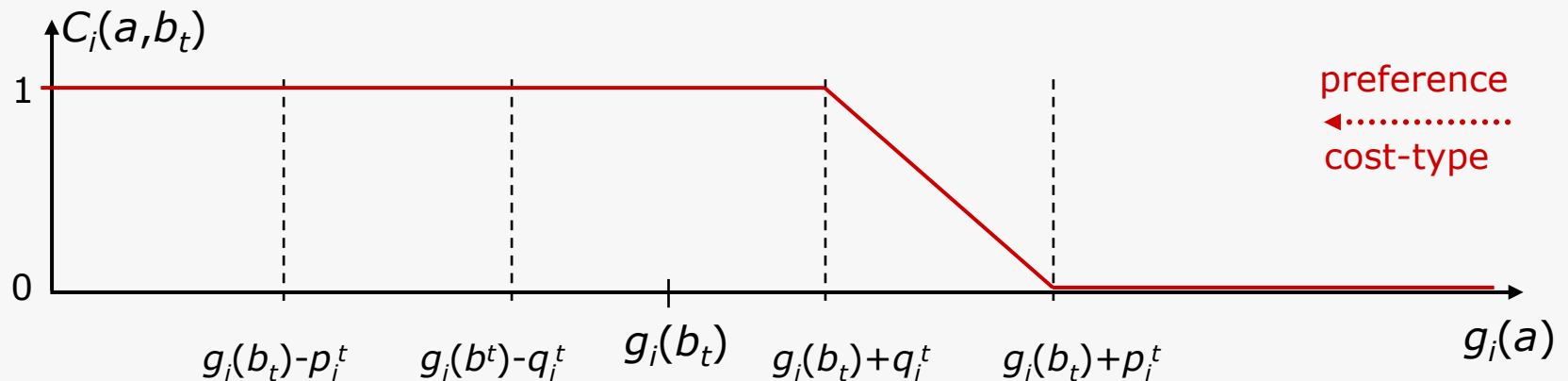
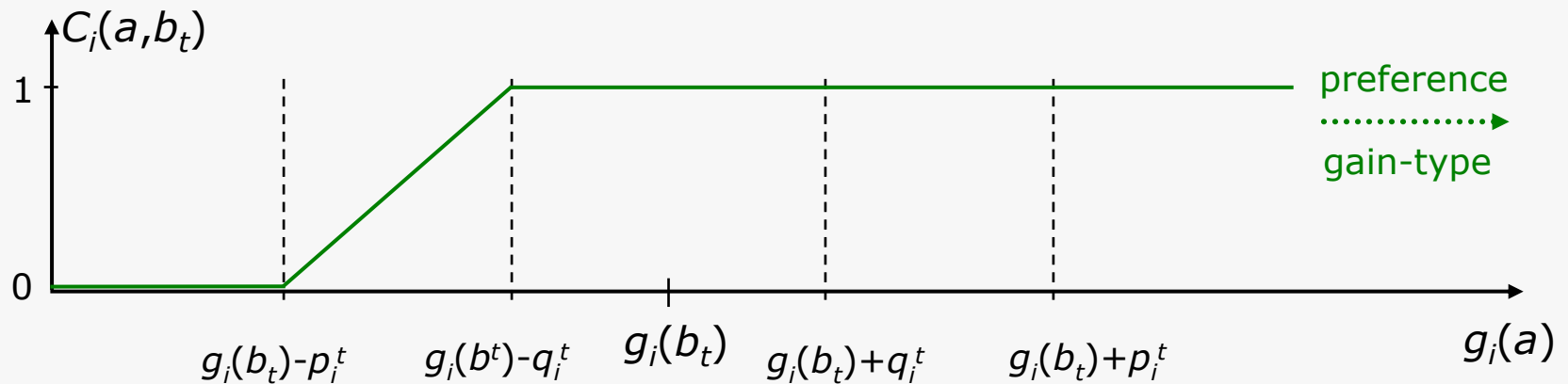
intercriteria: – importance coefficients (weights) of criteria  $k_i$   
– veto thresholds for each class limit  $v_i^t$ ,  $t=0,1,\dots,p$

- $0 \leq q_i^t \leq p_i^t \leq v_i^t$  are constant for each class limit  $b_t$ ,  $t=0,1,\dots,p$

- **Concordance and discordance tests** of ELECTRE TRI validate or invalidate the assertions  $aSb_t$  and  $b_tSa$

# ELECTRE TRI - concordance test

- Checks how strong is the coalition of criteria concordant with the hypothesis  $aSb_t$  (for each couple  $(a, b_t)$ ,  $\forall a \in A$  and  $t=0,1,\dots,p$ )
- Concordance coefficient  $C_i(a, b_t)$  for each criterion  $g_i$ :



## ELECTRE TRI - concordance test

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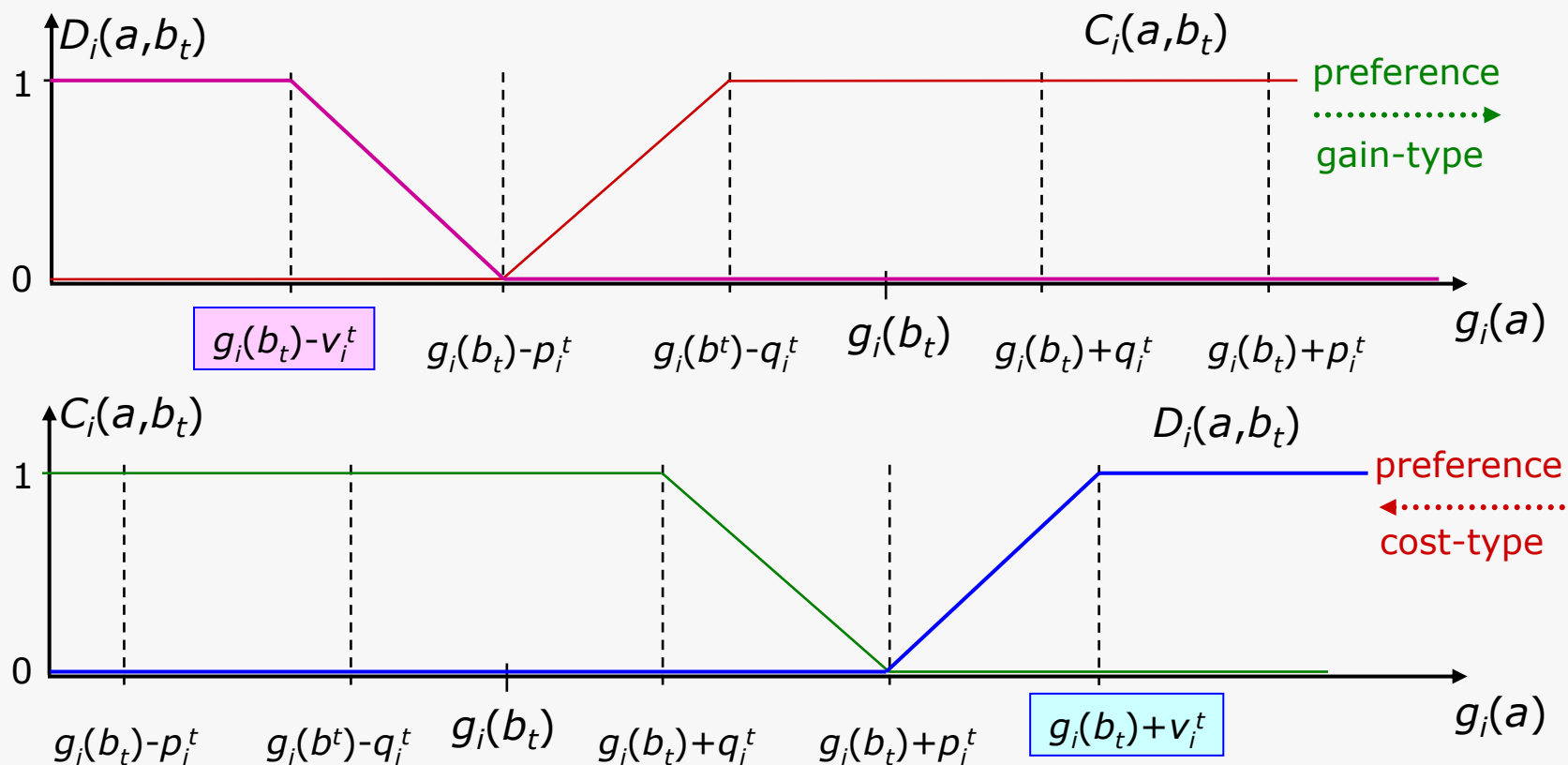
- Aggregation of concordance coefficients for  $a, b \in A$ :

$$C(a, b_t) = \frac{\sum_{i=1}^n k_i C_i(a, b_t)}{\sum_{i=1}^n k_i}$$

- $C(a, b_t) \in [0, 1]$

# ELECTRE TRI - discordance test

- Checks how strong is the coalition of criteria discordant with the hypothesis  $aSb_t$  (for each couple  $(a, b_t)$ ,  $\forall a \in A$  and  $t=0,1,\dots,p$ )
- **Discordance coefficient**  $D_i(a, b_t)$  for each criterion  $g_i$ :



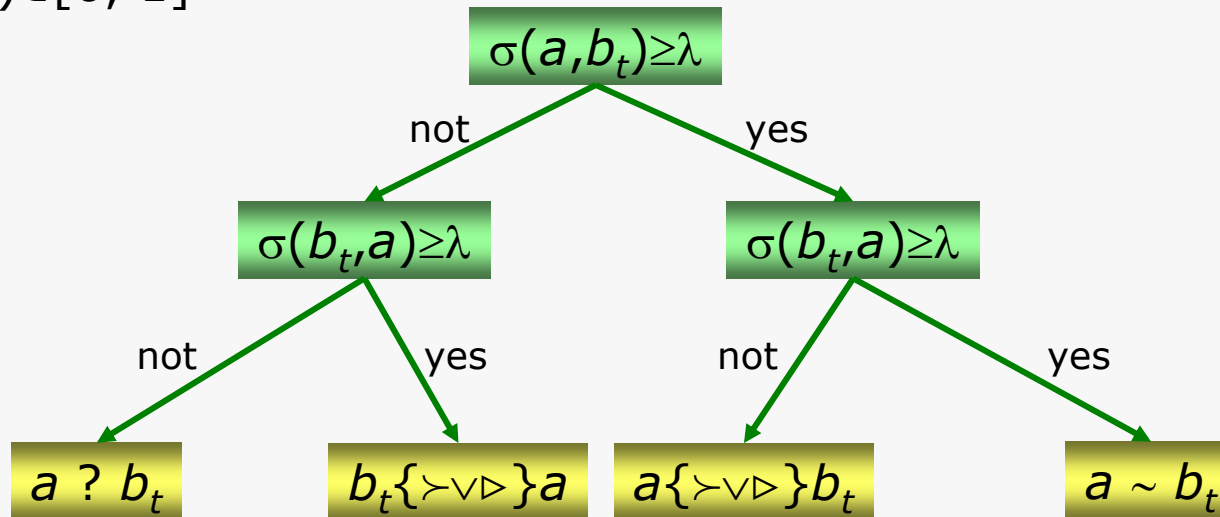
# ELECTRE TRI – credibility of the outranking relation

- Conclusion from concordance and discordance tests – credibility of the outranking relation:

$$\sigma(a, b_t) = C(a, b_t) \prod_{i \in F} \frac{1 - D_i(a, b_t)}{1 - C(a, b_t)}$$

where  $F = \{i : D_i(a, b_t) > C(a, b_t)\}$

- $\sigma(a, b_t) \in [0, 1]$

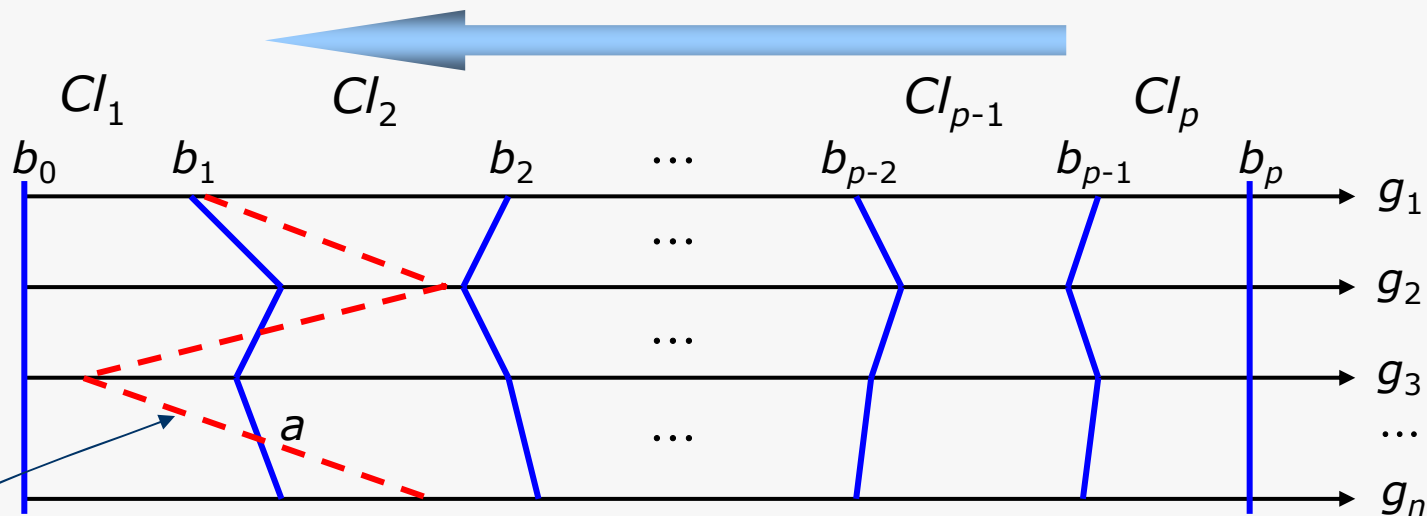


- What is the assignment of actions to decision classes ?

# ELECTRE TRI – exploitation of $S$ and final recommendation

- Assignment of actions to decision classes based on two ways of comparison of actions  $a \in A$  to limit profiles  $b_t$ ,  $t=0,1,\dots,p$ 
  - **Pessimistic**: compare action  $a$  successively to each profile  $b_t$ ,  $t=p-1,\dots,1,0$ ; if  $b_t$  is the first profile such that  $aSb_t$ , then  $a \rightarrow Cl_{t+1}$

comparison of action  $a$  to profiles  $b_t$

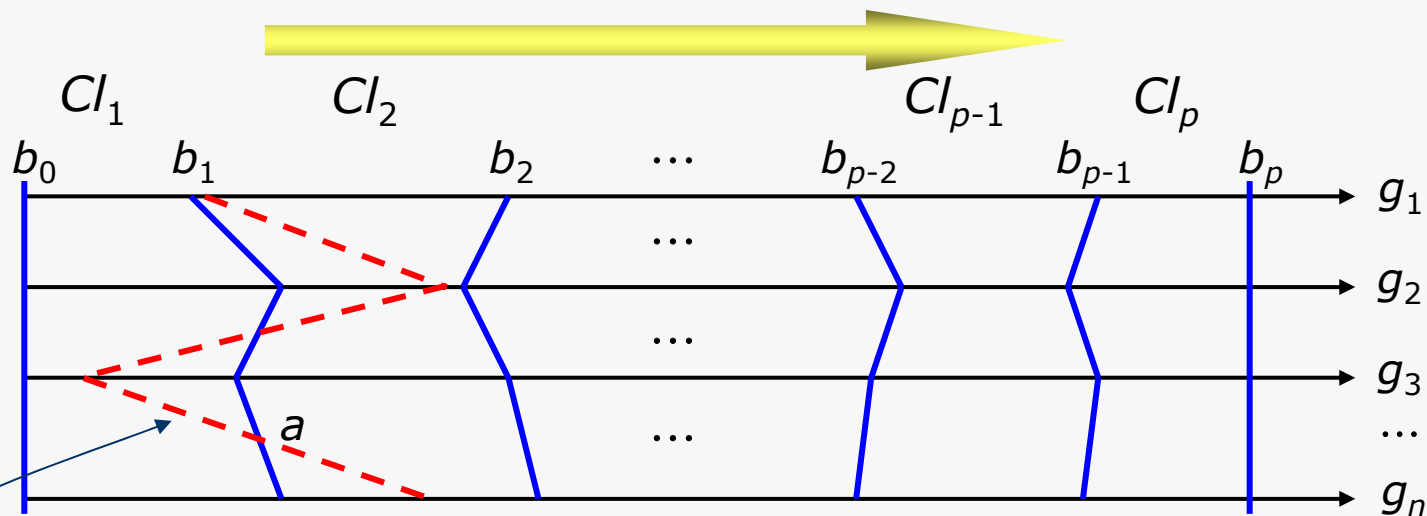


- **Pessimistic**, because for zero thresholds,  $a \rightarrow Cl_{t+1} \Leftrightarrow g_i(a) \geq g_i(b_t) \quad \forall i$ ,  
e.g.  $a \rightarrow Cl_1$

# ELECTRE TRI – exploitation of $S$ and final recommendation

- **Assignment of actions to decision classes** based on two ways of comparison of actions  $a \in A$  to limit profiles  $b_t$ ,  $t=0,1,\dots,p$ 
  - **Optimistic**: compare action  $a$  successively to each profile  $b_t$ ,  $t=1,\dots,p$ ; if  $b_t$  is the first profile such that  $b_t \{ \succ \vee \triangleright \} a$ , then  $a \rightarrow Cl_t$

comparison of action  $a$  to profiles  $b_t$



- **Optimistic**, because for zero thresholds,  $a \rightarrow Cl_t \Leftrightarrow g_i(b_t) < g_i(a) \forall i$ , e.g.  $a \rightarrow Cl_2$



# ELECTRE TRI - example

## Edit Project

Project

- Criteria
  - PRICE
  - TIME
  - COMFORT
- Profiles
  - M-G
  - B-M
- Alternatives
  - RER
  - METRO\_1
  - METRO\_2
  - BUS
  - TAXI
  - SNCF

Information Categories

Good

Category 3	Good
Category 2	Medium
Category 1	Bad

## Edit Project

Project

- Criteria
  - PRICE
  - TIME
  - COMFORT
- Profiles
  - M-G
  - B-M
- Alternatives
  - RER
  - METRO\_1
  - METRO\_2
  - BUS
  - TAXI
  - SNCF

Definition Performances Thresholds

Min ( $g_j$ ) = 2  
Max ( $g_j$ ) = 30  
Min ( $\Delta g_j$ ) = 1

PRICE	3
TIME	20
COMFORT	3

Indifference Threshold  
0

Preference Threshold  
1

Veto Threshold  
5

☐ Disable Veto



## Performances of Alternatives

	PRICE	TIME	COMFORT
RER	3	10	1
METRO_1	4	20	2
METRO_2	2	20	0
BUS	6	40	0
TAXI	30	30	3
SNCF	3	20	2

$$k_{\text{PRICE}}=3$$

$$k_{\text{TIME}}=5$$

$$k_{\text{COMFORT}}=2$$

$$\lambda=0.75$$

## Edit Project

Project

- Criteria
  - PRICE
  - TIME
  - COMFORT
- Profiles
  - M-G
  - B-M
- Alternatives
  - RER
  - METRO\_1
  - METRO\_2
  - BUS
  - TAXI
  - SNCF

Definition Performances Thresholds

Min ( $g_j$ ) = 2  
Max ( $g_j$ ) = 30  
Min ( $\Delta g_j$ ) = 1

PRICE	10
TIME	30
COMFORT	1

Indifference Threshold  
0

Preference Threshold  
1

Veto Threshold  
5

☐ Disable Veto

# ELECTRE TRI - example

Comparison to Profile		
	B-M	M-G
RER	>	>
METRO_1	>	<
METRO_2	>	<b>R</b>
BUS	<b>R</b>	<
TAXI	<	<
SNCF	>	<b>I</b>

Assignment by Category		
Category Name	Pessimistic Assignment	Optimistic Assignment
Good	RER	RER
Medium	SNCF	METRO_2
Bad		SNCF
Good	METRO_1	METRO_1
Medium	METRO_2	BUS
Bad		
Good	BUS	TAXI
Medium	TAXI	
Bad		

Statistics of Assignment		
Category Name	Pessimistic Assignment	Optimistic Assignment
Good	33 % (2 of 6)	50 % (3 of 6)
Medium	33 % (2 of 6)	33 % (2 of 6)
Bad	33 % (2 of 6)	17 % (1 of 6)

# ELECTRE TRI - example

Degrees of Credibility		
	B-M	M-G
RER	1.000 0.000	0.800 0.500
METRO_1	1.000 0.000	0.700 1.000
METRO_2	1.000 0.000	0.000 0.700
BUS	0.500 0.583	0.000 1.000
TAXI	0.000 0.800	0.000 1.000
SNCF	1.000 0.000	1.000 1.000

$$\lambda = 0.75$$

## Assignment by Alternative

Alternative Name	Pessimistic Assignment	Optimistic Assignment
RER	Good	Good
METRO_1	Medium	Medium
METRO_2	Medium	Good
BUS	Bad	Medium
TAXI	Bad	Bad
SNCF	Good	Good

