

DA2PL'20014 Paris, 21 November 2014

- 1. the usual imprecise knowledge of criteria significance weights, and
- 2. a small majority margin?





- 1. Modelling an uncertain criterion significance
- 2. Likelihood of "at least as good as" situations

Characterizing " at least as good as" situations Assessing the bipolar likelihood Examples

3. Confidence level of "outranking" situations

Outranking situations Confidence level Example

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Modelling uncertain criteria significances

We consider the criterion significance weight to be independent random variable W, distributing the potential significance weight of the given criterion around a mean value E(W) with variance V(W).

- 1. A continuous *uniform* distribution on the range 0 to 2 * E(W). Thus $W \sim U(0, 2E(W))$ and $V(W) = \frac{1}{3}E(W)^2$;
- 2. A symmetric beta(a, b) distribution with, for instance, parameters a = 2 and b = 2. Thus, $W \sim Beta(2, 2) \times 2E(W)$ and $V(W) = \frac{1}{5}E(W)^2$.
- 3. A symmetric *triangular* distribution on the same range with mode E(W). Thus $W \sim Tr(0, 2E(W), E(W))$ with $V(W) = \frac{1}{6}E(W)^2$;
- 4. A narrower beta(a, b) distribution with for instance parameters a = 4 and b = 4. Thus $W \sim Beta(4, 4) \times 2E(W)$, $V(W) = \frac{1}{9}E(W)^2$



•
$$A = \{x, y, z, ...\}$$
: a finite set of *n* potential decision actions;

- *F* = {1, ..., *n*}, a finite and coherent family of *m* performance criteria;
- [0; *M_j*]: Performance measurement scale used on criterion *j*;
- *ind_i*: Upper-closed indifference threshold;
- pr_j : Lower-closed preference threshold with $0 \leq ind_j < pr_j \leq M_j$;
- x_j: The marginal performance of any object x on criterion j;
- W_j : The random rational significance weight of criterion j.

$$r(\mathbf{x} \succeq_j \mathbf{y}) = \begin{cases} +1 & \text{if } x_j - y_j \ge -ind_j \\ -1 & \text{if } x_j - y_j \leqslant -pr_j \\ 0 & \text{otherwise.} \end{cases}$$
(1)

- +1 signifies x is performing at least as good as y on criterion j,
- -1 signifies that x is not performing at least as good as y on criterion j.
- 0 signifies that it is unclear whether, on criterion j, x is performing at least as good or not as y.





Performing globally "at least as good as"

Each criterion *j* contributes the random significance W_j of his marginal "at least as good as" characterization $r(\succeq_j)$ to the global characterization $\tilde{r}(\succeq)$ in the following way:

$$\widetilde{r}(\mathbf{x} \succeq \mathbf{y}) = \sum_{j \in F} \left[W_j \cdot r(\mathbf{x} \succeq_j \mathbf{y}) \right]$$
(2)

- $\tilde{r} > 0$ signifies x is globally performing at least as good as y,
- $\tilde{r} < 0$ signifies that x is not globally performing at least as good as y,
- $\tilde{r} = 0$ signifies that it is *unclear* whether x is globally performing at least as good or not as y.

Likelihood of "at least as good as" situations

From the Central Limit Theorem (CLT), we know that $\tilde{r}(x \geq y)$ (Eq. 2) leads, with *m* getting large, to a Gaussian variable Y with:

$$E(Y) = \sum_{j} E(W_{j}) \times r(x \succcurlyeq_{j} y),$$
$$V(Y) = \sum_{j} V(W_{j}) \times |r(x \succcurlyeq_{j} y)|.$$

Hence, the bipolar likelihood (*Ih*) of *validation*, respectively *invalidation* of a $(x \succeq y)$ situation may be assessed as follows:

$$lh(x \geq y) = 2 \times P(Y > 0.0) - 1.0 = -\operatorname{erf} \left(\frac{1}{\sqrt{2}} \frac{-E(Y)}{\sqrt{V(Y)}}\right).$$

The range of $lh(x \ge y)$ is [-1.0; +1.0], and $-lh(x \ge y) = lh(x \ge y)$, i.e. a negative value represents the likelihood of the negated outranking relation. A value +1.0 (resp. -1.0) means the outranking situation is certainly validated (resp. invalidated).



Example 1: equi-significant criteria

x and y are evaluated wrt 7 equi-significant criteria;

Four criteria positively support that x outranks y and three criteria support that x does not outrank y.

Suppose
$$E(W_j) = w$$
 for $j = 1, ..., 7$;
And $W_j \sim \mathcal{T}r(0, 2w, w)$ for $j = 1, ..., 7$;
Hence $E(\tilde{r}(x \geq y)) = 4w - 3w = w$,
And $V(\tilde{r}(x \geq y)) = 7 \times \frac{1}{6}w^2$.
If $w = 1$, $E[\tilde{r}(x \geq y)] = 1$ and $sd[\tilde{r}(x \geq y)] = 1.08$.
By the CLT, $lh(x \geq y) = 0.66 \approx 83\%$,
10 000 MC runs confirm $\tilde{r}(x \geq y) \rightsquigarrow Y = \mathcal{N}(1.03, 1, 089)$
with $P(Y \leq 0) \approx 17\%$.

Example 1 - continue

10000 simulations with 4 positive and 3 negative i.i.d. Tr(0,w,2w) weights





Example 2 - continue

Example 2: various significance weights

Table : Pairwise comparison of two decision alternatives

g j	g_1	g_2	g 3	g_4	g_5	g_6	g ₇		
$E(W_j)$	7	8	3	10	1	9	7		
a_1	14.1	71.4	87.9	38.7	26.5	93.0	37.2		
a ₂	64.0	87.5	67.0	82.2	80.8	80.8	10.6		
$a_1 - a_2$	-49.9	-16.1	+20.9	-43.5	-54.3	+12.2	26.5		
$r(\succcurlyeq_j)$	-1	0	+1	-1	-1	+1	+1		
F (~($\overline{\gamma}$								
E(r($a_1 \succcurlyeq a_2)$) =	$\sum_{i=1} r(a_1)$	≽ _j a ₁) ×	E(VV _j)				
= -7 + 0 + 3 - 10 - 1 + 9 + 7 = +1									

If now $W_j \sim \mathcal{T}r(0, 2E(W_j), E(W_j))$, how confident can we be about the actual positiveness of $\tilde{r}(a_1 \geq a_2)$?



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The confident outranking relation \succeq

From an epistemic point of view, we say that:

- 1. action x outranks action y, denoted $(x \succeq y)$, if
 - 1.1 a confident majority of criteria validates a global outranking situation between x and y, and
 - 1.2 no veto is observed on a discordant criterion,
- 2. action x does not outrank action y, denoted $(x \not\gtrsim y)$, if
 - 2.1 a confident majority of criteria invalidates a global outranking situation between x and y, and
 - 2.2 no counter-veto is observed on a concordant criterion.



Considerably better or worse performing situations

On a criterion j, we characterize a *considerably less performing* situation, called veto and denoted \ll_i , as follows:

$$r(\mathbf{x} \ll \mathbf{y}) = \begin{cases} +1 & \text{if } x_j + v_j \leq y_j \\ -1 & \text{if } x_j - v_j \geq y_j \\ 0 & \text{otherwise.} \end{cases}$$
(3)

where v_j represents a veto discrimination threshold. A corresponding dual *considerably better performing* situation, called counter-veto and denoted \gg_j , is similarly characterized as:

$$r(\mathbf{x} \gg_j \mathbf{y}) = \begin{cases} +1 & \text{if } x_j - v_j \ge y_j \\ -1 & \text{if } x_j + v_j \le y_j \\ 0 & \text{otherwise.} \end{cases}$$
(4)

Veto and counter-veto situations

A global considerable worst performaning (*veto*) situation, or considerably better perform ing (*counter-veto*) situation is now defined as follows:

$$r(\mathbf{x} \ll \mathbf{y}) = \bigotimes_{j \in F} r(\mathbf{x} \ll_j \mathbf{y})$$
 (5)

$$r(x \gg y) = \bigotimes_{j \in F} r(x \gg_j y)$$
 (6)

where \bigcirc represents the epistemic polarising (Bisdorff 1997) or symmetric maximum (Grabisch et al. 2009) operator:

$$r \otimes r' = \begin{cases} \max(r, r') & \text{if } r \ge 0 \land r' \ge 0, \\ \min(r, r') & \text{if } r \le 0 \land r' \le 0, \\ 0 & \text{otherwise.} \end{cases}$$
(7)

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Characterizing veto and counter-veto situations

- 1. $r(x \ll y) = 1$ iff there exists a criterion j such that $r(x \ll_j y) = 1$ and there does not exist otherwise any criterion k such that $r(x \gg_k y) = 1$.
- 2. Conversely, $r(x \gg y) = 1$ iff there exists a criterion j such that $r(x \gg_j y) = 1$ and there does not exist otherwise any criterion k such that $r(x \ll_k y) = 1$.
- 3. $r(x \gg y) = 0$ if either we observe no very large performance differences or we observe at the same time, both a very large positive and a very large negative performance difference.

Comment

$$r(\not\ll)^{-1}$$
 is identical to $r(\gg)$.

Polarising the global "at least as good as " characteristic

The outranking characteristic $\tilde{r}(\succeq)$ is defined as follows:

$$\widetilde{r}(x \succeq y) = \left[\, \widetilde{r}(x \succcurlyeq y) \odot - r(x \lll y) \, \right]$$

And in particular,

- 1. $\tilde{r}(x \succeq y) = \tilde{r}(x \succcurlyeq y)$ if no very large positive or negative performance differences are observed,
- 2. $\tilde{r}(x \succeq y) = 1$ if $\tilde{r}(x \succcurlyeq y) \ge 0$ and $r(x \ggg y) = 1$,
- 3. $r(x \succeq y) = -1$ if $\tilde{r}(x \succcurlyeq y) \leq 0$ and $r(x \ll y) = 1$,
- 4. $\tilde{r}(x \succeq y) = 0$ in all other cases, and especially if conjointly $r(x \ggg y) = 1$ and $r(x \lll y) = 1$.

Confidence level α for outranking situations

By requiring now a certain level α of likelihood for confidently validating all pairwise outranking situations, we may thus enforce the actual confidence we may have in the valued outranking digraph.

For any outranking situation $(x \succeq y)$ we obtain:

$$\hat{r}_{\alpha}(x \succeq y) = \begin{cases} E[\tilde{r}(x \succeq y)] & \text{if } \operatorname{abs}(lh(x \succcurlyeq y)) \geqslant \alpha, \\ 0 & \text{otherwise.} \end{cases}$$
(8)

If $E(W_j) = w_j$, $E[\tilde{r}(x \succeq y)]$ equals the corresponding deterministic outranking characteristic $r(x \succeq y)$. We safely preserve, hence, in our stochastic modelling, all the nice structural properties of the deterministic outranking relation like

weak completeness and coduality.

Example 3: Confident outranking digraph

gi	Wi	a ₁	a ₂	a ₃	a4	<i>a</i> 5	a ₆	a ₇
g_1	7	14.1	64.0	73.4	36.4	30.6	85.9	97.8
g ₂	8	71.4	87.5	61.9	84.7	60.4	54.5	45.8
g ₃	3	87.9	67.0	25.2	34.2	87.3	43.1	30.4
g ₄	10	38.7	82.2	94.1	86.1	34.1	97.2	72.2
g_5	1	26.5	80.8	71.9	21.3	56.4	88.1	15.0
g_6	9	93.0	80.8	23.2	57.2	81.4	16.6	93.0
g7	7	37.2	10.6	64.8	98.9	69.9	24.7	13.6
Τł	hresho	olds: ind	$d_i = 10.$	0, $pr_i =$	20, and	d $v_i = 8$	80 for i	∈ <i>F</i> .

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Example 3: Confident outranking digraph

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Table : Deterministic credibility of $(x \succeq y)$

$r(\succeq) imes 45$	a ₁	a ₂	a ₃	a4	<i>a</i> 5	a ₆	a ₇
a ₁	-	+1	-5	-11	+22	+9	0
a ₂	+16	-	+21	0	+25	+14	+22
a ₃	+21	+5	-	-3	+21	+34	+13
a ₄	+21	+45	+29	-	+19	+19	+45
a_5	+28	-7	+10	-5	-	+9	+2
<i>a</i> ₆	+6	+5	+31	-3	+7	-	+20
a ₇	+45	+11	+1	0	+15	+13	-

Table : CLT likelihood of the $(x \geq y)$ situations

lh	a ₁	a ₂	a ₃	a ₄	<i>a</i> 5	<i>a</i> ₆	a ₇
a ₁	-	+.11	49	89	+1.0	+.76	+.85
<i>a</i> ₂	+.98	-	+1.0	+1.0	+1.0	+.98	+1.0
a ₃	+.99	+.49	-	30	+.99	+1.0	+.91
a_4	+.99	+. 49	+1.0	-	+.99	+1.0	+.91
a_5	+1.0	64	+.81	49	-	+.76	+.23
a_6	+.66	+.49	+1.0	30	+.64	-	+1.0
a ₇	+.70	+.91	+.11	+.56	+.97	+.94	-

Motivation Modelling an uncertain criterion significance	Likelihood of " <i>at least as good as</i> " situations ooo oooo	Confidence level of "outran 00000 0000	Motivation Modelling an uncertain criterion significance	Likelihood of " <i>at least as good as</i> " situations ooo oooo	Confidence level of "ou 00000 0 0000
Example 3: Confi	dent outranking digra	aph	(Content	
			1. Modelling an uncertain cri	terion significance	
Table : 90% confi	ident outranking situations		2. Likelihood of " <i>at least as</i> Characterizing " <i>at leas</i>	good as" situations t as good as" situations	

$\hat{r}_{90\%}(x \succeq y)$	a_1	a ₂	a 3	a 4	a 5	a 6	a 7
a_1	-	0 (+1)	0 (-5)	-11	+22	0 (+9)	0
a_2	+16	-	+21	0	+25	+14	+22
a 3	+21	0 (+5)	-	0 (-3)	+21	+34	+13
a_4	+21	0 (+45)	+29	-	+19	+19	+45
a_5	+28	0 (-7)	+10	0 (-5)	-	0 (+9)	0 (+2)
a_6	0 (-7)	0 (+5)	+31	0 (-3)	0 (+7)	-	+20
a ₇	0 (+45)	+11	0 (+1)	0	+15	+13	-

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Exploiting the confident outranking digraph

Table : Pairwise comparison of alternatives a_4 and a_2

gj	g ₁	<i>g</i> 2	<i>g</i> 3	<i>g</i> 4	<i>g</i> 5	<i>g</i> 6	g ₇
Wj	7	8	3	10	1	9	7
а _{4j}	36.5	84.7	34.2	86.1	21.3	57.2	<mark>98.9</mark>
а _{2j}	60.0	87.5	67.0	82.2	<mark>80.8</mark>	<mark>80.8</mark>	10.6
$egin{aligned} &(a_{4j}-a_{2j})\ &r(a_4 \succcurlyeq_j a_2)\ &r(a_4 \ll _j a_2)\ &r(a_4 \gg _j a_2)\ &r(a_4 \gg _j a_2) \end{aligned}$	-27.5 -1 0 0	-2.8 +1 0 0	-32.8 -1 0 0	+3.8 +1 0 0 0	-59.2 -1 0 0	-23.6 -1 0 0	+88.8 +1 0 +1

Thresholds: $ind_j = 10.0$, $pr_j = 20$, and $v_j = 80$ for $j \in F$.

$$\tilde{r}(a_4 \succcurlyeq a_2) = +5 \text{ and } \tilde{r}(a_4 \succsim a_2) = +45.$$

Yet, $lh(a_4 \succcurlyeq a_2) = 0.49 < 0.80$, hence
 $\hat{r}_{.80}(a_4 \succsim a_2) = 0.$

The confident outranking digraph

The deterministic digraph:



The 90% confident digraph:





Rubis Python Server (graphyiz), R. Bisdorff, 200



- We illustrate some simple models for tackling uncertain significance weights: uniform, triangular and beta laws.
- Applying the Central Limit Theorem, we are able to compute the actual likelihood of any pairwise *at least as good as* and *not at least as good as* situations.
- This operational result allows to enforce a given confidence level on the corresponding outranking situations.
- On a small illustrative best choice problem, we eventually show the pragmatic decision aid benefit one may expect from exploiting a confident versus a classic deterministic outranking digraph.

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