

**MEASURE OF INCONSISTENCY. AHP & POTENTIAL
MEETHOD. A COMPARISON.**

LAVOSLAV ČAKLOVIĆ

PREFERENCE GRAPH

A **preference graph** is a digraph $\mathcal{G} = (V, \mathcal{A})$ where V is a set of nodes and \mathcal{A} is a set of arcs of \mathcal{G} . We say that the node a is **more preferred** than the node b , in notation $a \succ b$, if there is an arc (a, b) outgoing from b and ingoing to a . A **preference flow** is a non-negative real function \mathcal{F} defined on the set of arcs. The value \mathcal{F}_α on the arc α is an intensity of the preference on some scale¹. For the arc $\alpha = (a, b)$, $\mathcal{F}_\alpha = 0$ means that the decision maker is indifferent for the pair $\{a, b\}$. In that case orientation of the arc is arbitrary. A preference flow is **consistent** if there is no component of the flow in the cycle-space of the graph. According to that definition, \mathcal{F} is consistent if the sum of algebraic components of the flow along each cycle is equal to zero. Equivalently, \mathcal{F} is consistent if there exists $X \in \mathbb{R}^n$ (called **potential**) such that $AX = \mathcal{F}$, where A is an incidence matrix of the graph. If \mathcal{F} is not consistent, the potential X may be calculated as a solution of the equation $AX = \mathcal{F}_0$ where \mathcal{F}_0 is the best approximation of \mathcal{F} in the column space of the matrix A . It is evident that a potential X of the consistent flow is a measurable value function on the set of nodes, i.e.

$$\begin{aligned}\mathcal{F}_{(a,b)} \geq 0 &\iff X(a) \geq X(b) \\ \mathcal{F}_{(a,b)} \geq \mathcal{F}_{(c,d)} &\iff X(a) - X(b) \geq X(c) - X(d).\end{aligned}$$

In Saaty's Eigenvalue Method (EVM) input data are captured in a positive reciprocal matrix $A = (a_{ij})$. By definition, A is consistent if

$$a_{ij}a_{jk} = a_{ik}, \quad i, j, k = 1, \dots, n.$$

A connection between those two types of consistency is the following: If we define a flow by

$$\mathcal{F}_{(i,j)} := \log(a_{ij}).$$

then, \mathcal{F} is consistent iff A is consistent.

RANK REVERSAL AND RANDOMIZATION.

We say that a **condition of order preservation** (COP) for AHP is satisfied if

$$a_{ij} > a_{kl} \implies \frac{w_i}{w_j} > \frac{w_k}{w_l},$$

where $A = (a_{ij})$ is the reciprocal positive matrix and w its Perron eigenvector. For the preference flow \mathcal{F} , we say that COP is satisfied if:

$$\mathcal{F}_{(i,j)} > \mathcal{F}_{(k,l)} \implies X_i - X_j > X_k - X_l.$$

¹For subjective pairwise comparisons the scale is $\{0, 1, 2, 3, 4\}$.

A measure of inconsistency in EVM is given by *inconsistency index* (CI), in Potential Method (PM) inconsistency is measured by an *angle* between \mathcal{F} and \mathcal{F}_0 measured in degrees (DEG). To compare those measures we performed 10^4 simulations of positive reciprocal matrix. For each randomly generated reciprocal matrix we calculate its inconsistency index CI the inconsistency measure DEG of the corresponding flow and a number of violations of COP (NOV) for EVM and PM respectively.

CONCLUSION.

It is shown that NOV and DEG are very good correlated for PM ($r = 0.811$) and NOV and CI are not good correlated for EVM ($r = 0.460$) and NOV for both methods are highly correlated ($r = 0.951$), which allows to conclude that inconsistency measure for EVM is not well designed and it should be used by caution. It is also shown that DEG is distributed as a Gumbel Distribution. For instance, if the randomization is made as a log-normal perturbation $N(0, 1)$ of the random consistent flow, the inconsistency measure DEG is distributed as the Gumbel Distribution $E(\alpha = 17.61, \beta = 7.03)$. This allows to define an *upper bound* for admissible inconsistency of DEG as a p -quantile ($p = 0.05$) of the random degree distribution as a function of the number of nodes in the graph.

TABLE 1. Quantiles of random degree as a function of the nodes number. 10^5 simulations.

nodes number	perturbation ($\sigma = 1$)	Gumbel Distribution $E(\alpha, \beta)$			
		0.05-quantile		α	β
		from data	theoretical		
3	normal	1	-3.1	9.43	11.41
	uniform	1	-2.8	9.81	11.48
4	normal	6	5.3	15.01	8.83
	uniform	7	5.9	15.43	8.66
5	normal	10	9.9	17.61	7.03
	uniform	11	10.6	18.09	6.81
6	normal	13	12.7	19.18	5.91
	uniform	13	13.4	19.59	5.6
7	normal	15	14.7	20.24	5.07
	uniform	15	15.3	20.59	4.80
8	normal	16	16.1	21.03	4.47
	uniform	17	16.7	21.34	4.22
9	normal	17	17.2	21.64	4.02
	uniform	18	17.7	21.88	3.79
10	normal	18	18.0	22.06	3.67
	uniform	18	18.5	22.31	3.44

In randomization procedure we used *Perl* and data analysis was performed by *Mathematica* and *R*.