

AN INTEGRATED AHP AND WEIGHTED FUZZY GOAL PROGRAMMING MODEL FOR IS PROJECT SELECTION

ABSTRACT

The purpose of this paper is to develop an integrated AHP and Fuzzy Goal Programming methodology which deals the imprecise data and offer more flexibility. The proposed method includes the following steps: At first, an expert team is formed which identifies the decision criteria and alternatives and builds a hierarchical model for IS project selection. After that, the AHP is used in order to obtain weights of each criterion and project. At the end, a Weighted Additive Fuzzy Goal Programming model (WAFGP) is formulated and used to complete the project selection decision. In order to illustrate the use and advantages of this approach; a hypothetical example has been exposed. The results show the quality of the support which the proposed model provide to the IS project selection decision. Despite its advantages, the methodology proposed here neglects the uncertain nature of decision maker's judgment and the interdependencies among criteria and alternatives.

Keywords: Information systems, project selection, Analytic hierarchy process (AHP), Weighted Additive Fuzzy Goal Programming.

1. Introduction

Us a multi-criteria decision making, IS projects selection has to consider a large number of alternatives, multiple and often conflicting, fuzzy and imprecise attributes, and also interdependencies among this alternatives and criteria. However, if the existing methods reported in literature consider the diversity of the attributes and the interdependencies problems, they neglects the imprecise and fuzzy nature of this attributes. Furthermore, the Preemptive/Lexicographic Goal Programming used is not flexible when dealing with integer problem with many goals. This paper proposes a hybrid method using AHP and Weighted Additive Fuzzy zero-one goal programming in order to deals with this limits.

2. Literature Review

During the last decades, several methodologies have been developed to overcome the IS project selection difficulties. As for example, Schniederjans & Wilson (1991) have proposed an AHP- zero-one linear programming methods to consider budgetary and resource constraints. After that, Lee & Kim (2001) have developed an integrated Delphi-ANP-ZOGP models to estimate the degree of interdependencies among IS projects. Finally, to deal with the imprecise data in IS projects and uncertain judgment of decision makers; Bolat and al. (2014) have developed an integrated fuzzy AHP – fuzzy multi-objective linear programming model. Beyond the progresses, some weaknesses still characterize the IS project selection methodologies. Actually, the Preemptive/Lexicographic Goal Programming (with priority) used in almost of the studies is not flexible when dealing with integer problem with many goals.

3. Objectives

The aim objective of this paper is to present a hybrid method using the analytic hierarchy process (AHP) proposed by Saaty (1980) and the Weighted Additive Fuzzy Goal Programing (WAFGP) proposed by Yaghoobi & al. (2008) to offer more flexibility.

4. Research Design/Methodology

The proposed methodology includes the following steps: At first, the decision criteria and alternatives are identified and a hierarchical model for IS project selection is constructed.

After that, the AHP is used in order to obtain weight of each criterion. At the end, a WAFGP model is formulated and used to complete the project selection decision.

5. An Illustrative Application

In order to illustrate the use and advantages of the proposed methodology, a hypothetical example is described as follows. Suppose that a firm have to choice 05 information system projects among 10 alternatives. For this decision, 11 criteria that appear on the figure 01 have been considered.

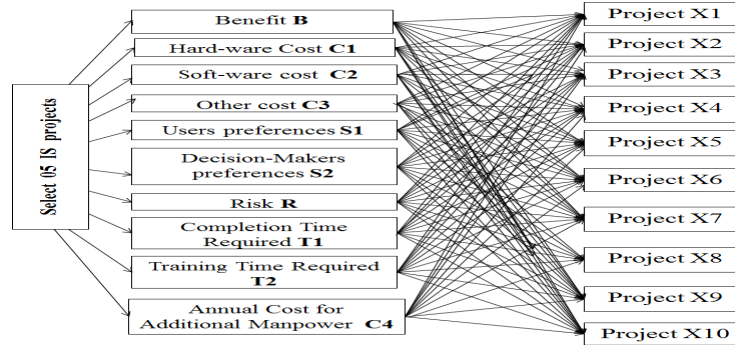


Fig 1. AHP hierarchy

At first, the hierarchy model exposed in the figure 01 has been constructed and the attributes and alternatives have been presented to the project team members to extract judgment matrices with a nine-point scale at each level. The software SUPERDECISION have been then used to determine the normalized weights and synthesize the results. Table 01 lists the Pairwise Comparison Judgment Matrices and the relative weights of attributes. The decision makers were fairly consistent in ranking the attributes. Indeed, the consistency indexes have been less than the threshold value 0.1 (Saaty, 1980).

Table 01: Pairwise comparison judgment matrices and relative weights of attributes

	C1	C2	C3	C4	B	R	T1	T2	S1	S2
C1	1	1	3	3	2	2	3	4	3	3
C2	1	1	3	3	2	2	3	4	3	3
C3	1/3	1/3	1	2	1/2	1/3	2	3	3	3
C4	1/3	1/3	1/2	1	1/2	1/3	3	2	3	3
B	1/2	1/2	2	2	1	2	2	3	4	4
R	1/2	1/2	3	3	1/2	1	2	3	4	4
T1	1/3	1/3	1/2	1/3	1/2	1/2	1	2	3	3
T2	1/4	1/4	1/3	1/2	1/3	1/3	1/2	1	4	4
S1	1/3	1/3	1/3	1/3	1/4	1/4	1/3	1/4	1	1/2
S2	1/3	1/3	1/3	1/3	1/4	1/4	1/3	1/4	2	1
*W	0.18798	0.18798	0.09023	0.07404	0.13747	0.13392	0.06653	0.05591	0.03074	0.03521

* Relative Weights of Attributes. Inconsistency: 0.06063.

In the third step, the type and the data of membership functions used for each objective have been determined (Table 03) and the WAFGP model has been formulated as follows:

$$\begin{aligned}
 \text{Min } Z = & 0,13747 \left[\frac{n_1}{10000} \right] + 0,18798 \left[\frac{p_2}{20000} \right] + 0,18798 \left[\frac{p_3}{10000} \right] + 0,09023 \left[\frac{p_4}{300} \right] + 0,13392 \left[\frac{p_5}{200000} \right] \\
 & + 0,03521 \left[\frac{n_6}{10} \right] + 0,03074 \left[\frac{n_7}{10} \right] + 0,06653 \left[\frac{p_8}{2000} \right] + 0,05591 \left[\frac{p_9}{3000} \right] + 0,07404 \left[\frac{p_{10}}{100} \right] \\
 \text{subject to :} &
 \end{aligned}$$

$$\begin{array}{llll}
 \sum_{i=1}^{10} B_i x_i + n_1 \geq 480 & \sum_{i=1}^{10} PRD_i x_i + n_6 \geq 47 & u_2 + \frac{1}{20000} = 1 & u_8 + \frac{1}{2000} = 1 \\
 \sum_{i=1}^{10} h_i x_i - p_2 \leq 650 & \sum_{i=1}^{10} PRU_i x_i + n_7 \geq 49 & u_3 + \frac{1}{10000} = 1 & u_9 + \frac{1}{3000} = 1 \\
 \sum_{i=1}^{10} S_i x_i - p_3 \leq 280 & \sum_{i=1}^{10} t_i x_i - p_8 \leq 0 & u_4 + \frac{1}{300} = 1 & u_{10} + \frac{1}{100} = 1 \\
 \sum_{i=1}^{10} O_i x_i - p_4 \leq 360 & \sum_{i=1}^{10} tt_i x_i - p_9 \leq 0 & u_5 + \frac{1}{200000} = 1 & x_1 = 1 \\
 \sum_{i=1}^{10} r_i b_i x_i - p_5 \leq 0 & \sum_{i=1}^{10} m_i x_i - p_{10} \leq 11 & u_6 + \frac{1}{10} = 1 & \sum_{i=1}^{10} x_i = 5 \\
 & & u_7 + \frac{1}{10} = 1 & x_i = 0 \text{ or } 1 \\
 & & & i = 1, \dots, 10 \\
 & u_1 + \frac{1}{10000} = 1 & &
 \end{array}$$

Where : B_i = the benefit derived from implementing project i , h_i = the hardware cost associated with implementing project i , S_i = the software cost associated with implementing project i , O_i = the other costs associated with implementing project i , r_i = the likelihood of failure of project i ; PRD_i = the decision-maker's preference for project i , PRU_i = the user's preference for project i , t_i = the estimated completion time for project i , tt_i = the estimated training time required for project i , m_i = the cost of additional manpower for project i , p_i, n_i = the positive and negative deviation variables for the goals i , μ_i = degree of membership functions for the goal i , $i = 1, 2, \dots, n$ IS project goals.

In this model, x_i is a binary variable so that it takes the value of 1 if the project i is selected, it takes the value 0 otherwise. Using the LINGO package, the obtained optimal solution is as follows: $x_1 = 1$, $x_2 = 0$, $x_3 = 1$, $x_4 = 0$, $x_5 = 0$, $x_6 = 0$, $x_7 = 0$, $x_8 = 1$, $x_9 = 1$, $x_{10} = 1$. The proposed model determines degree of membership functions for the i th goal:

$$(\mu_1, \mu_2, \mu_3, \mu_4, \mu_5, \mu_6, \mu_7, \mu_8, \mu_9, \mu_{10}) = (0.87, 1, 1, 1, 0.341, 0.11, 0.53, 0.60, 0.84, 0.89, 0.93)$$

6. Limitations

Despite its advantages, the methodology proposed here neglected the uncertain nature of decision maker's judgment and the interdependencies among IS projects.

7. Conclusions

The IS project selection approach offered here combines the AHP within the WAFGP. The AHP is first used to estimate the decision criteria weights. These weights are used to formulate a WAFGP model and to complete the project selection decision. In comparison with the previews methodologies, this on deals with the imprecise data in IS projects and seems to be easier and simpler. Despite its advantages, this approach neglects the uncertain nature of decision maker's judgment and the interdependencies among IS projects. To overcome this limits, a hybrid models using fuzzy ANP and fuzzy parameters could be developed in the future.

8. Key References

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9. Appendices

Table 02: The model data inputs

Project	Mandated	Benefit*	Hardware Cost*	Software cost*	Other Cost*	Decision-Makers preferences	Users preferences	Risk factor	Completion Time**	Training Time**	Annual Cost for Additional Manpower*
1	Yes	1774	1900	3800	00	9.336	9.762	3	50	90	500
2	No	1349	11500	2254	160	9.305	9.638	3	43	18	286
3	No	40600	29500	16020	00	9.349	9.773	4	90	19	545
4	No	1200	21000	7800	18	7.727	8.008	3	60	66	29
5	No	5000	20000	750	190	9.272	9.505	2	83	84	294
6	No	3000	14000	44	20	8.661	9.517	2	67	136	100
7	No	2090	320	16000	00	9.206	9.377	3	91	69	00
8	No	1300	500	1000	30	8.604	9.286	3	97	119	00
9	No	1320	1200	3300	08	7.552	8.193	2	28	61	39
10	no	1720	00	2500	10	7.481	8.002	2	36	24	23
Maximum available	-----	48000	65000	28000	360	47	49	-----	-----	-----	1100

* In 1000\$.

** Required in days.

Table 03: Type and data of memberships function for every goal

objective	Type of membership functions	data of membership functions	
Benefit-related objective	Type 2	(Δ_{iL}, b_i)	(10000, 48000)
Hardware cost-related objective	Type 1	(b_i, Δ_{iR})	(65000, 20000)
Software cost-related objective	Type 1	(b_i, Δ_{iR})	(28000, 10000)
Other cost-related objective	Type 1	(b_i, Δ_{iR})	(360, 300)
Risk-related objective	Type 1	(b_i, Δ_{iR})	(0, 200000)
Preference-related objectives	Type 2	(Δ_{iL}, b_i)	(10, 47)
Preference-related objectives	Type 2	(Δ_{iL}, b_i)	(10, 49)
Completion time required	Type 1	(b_i, Δ_{iR})	(0, 2000)
Training time required	Type 1	(b_i, Δ_{iR})	(0, 3000)
Additional manpower required	Type 1	(b_i, Δ_{iR})	(1100, 100)