Original Article

Supporting Structure for Prioritizing and Assigning Six Sigma Software Projects Using Fuzzy Logic Topsis and Fuzzy Logic Expert System

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Abstract – The software project selection process system can be encountered as a very important action in the success of six sigma software projects. Such a manner that is prioritizing and assigning software projects to execute software development teams is recognized, the very important step in software process plentiful research studies has been undergoing into six sigma software project selection, no one has been concentrate on selecting and assigning software projects as an articulate process occurring simultaneously. In this concern, this paper presents a supporting structure for decision making choosing and allocating the six sigma software projects to execute software project development teams, the main important action of the six sigma software project selection process is selected. Coming after, recognizing six sigma potential software projects in the small and medium scale industries. The fuzzy logic TOPSIS methodology is used to rank them. Further, the strong effect and effort indices for every software project are computed. Afterwords fuzzy logic expert system is used to assign the software projects to six sigma champions. Finally, a case study in the software organizations is presented, and the supporting structure is discussed to illustrate its software developed application.

Keywords – Six Sigma Project Selection, Fuzzy Logic Expert System, TOPSIS, Software Project Allocation, Multiple Criteria Decision Making.

I. INTRODUCTION

In order to execute software projects using six sigma methodologies named DMAIC (Define, Measure, Analyze, Improve, and Control), this methodology is considered a very important key to achieving software project goals [1][18]. Now a day's, technology demonstrates the present software industry, covering the way for the construction of dynamic environments [2][19]. Software companies require continuous improvements in order to survive fierce completion [3][20]. The six sigma methodology concentrates on the most famous lean six sigma tools for waste elimination, expenditure reduction and improvement of quality throughout the organizations [4][21]. And to continuously define and solve several problems. This paper discussed the prioritizing of six sigma software projects, but there is no other model developed for assigning software projects to implement teams and six sigma champions. This is the best model for prioritizing and choosing six sigma software projects and also assigns them to corresponding champions simultaneously [22][5]. Therefore, this article is innovative in a serious attempt to achieve something to develop a comprehensive supporting structure, presenting a step by step software process to identify potential six sigma software projects throughout an s/w industry, prioritize potential software projects using fuzzy logic TOPSIS, and assigning them to champions, according to their proficiencies, through a fuzzy logic methodology, at that time [23][7].

II. LITERATURE REVIEW

A. Six sigma software project selection criteria

In the year of 1987, Motorola company was introduced the six sigma methodology, and it was utilized by electric components in early 1995. Its intention was to produce long term defect levels, DPMO [Defects Per Million Opportunities] was below 3.4σ [14][15][24]. Using six sigma improve the creative activities and produces high-level outputs, enhancing workers skills, improving operational processes, and encouraging changes [25][26][27]. Six Sigma methodologies is a technical tool for creating value. This is for achieving business success [28]. It wrongly selected six sigma software projects beginning; they will be incapable and lead to a loss in time and costs [29][30].

According to research carried out in this field, a set of criteria that can be utilized in choosing six sigma software projects are known and presented as follows:

- Software customer satisfaction
- Investment returns
- Software project scope
- Availability of resources, including Manpower
- Time requirement
- Satisfaction of software developers from software projects
- Increasing profits and financial efficiency.

B. Six sigma potential software projects identification

In potential software project selection to collect information from various sources, in this regard, the very important resources are:

- Software customer satisfaction estimation results
- Software customers voice
- Software suppliers
- Software analysis competitor
- Software cost of quality and reliability reports.
- Software waste identification
- Software previous projects

The six sigma software projects can be identified in every software industry by collecting software information from the brainstorming meeting.

C. Fuzzy logic TOPSIS model

First, Hwang and Yoon proposed a model as the TOPSIS model. This is one of the multiple attribute decision making models. This is used for selecting the best alternative and selecting software projects. The TOPSIS model is considered a major multiple attribute decision-making method in comparison with other related methods like the analytical hierarchical approach.

• TOPSIS can consist of an unrestricted level of software criteria and software performance properties.

• TOPSIS can have changes in one property that can be neutralized by other properties in an indirect or direct manner.

• TOPSIS provides us with a systematic procedure, streamlined with a relatively simple evaluation process.

• TOPSIS avoids Pairwise comparisons needed by techniques such as hierarchical analytical process.

• TOPSIS not only provides us with a preferential ranking of alternatives but also computes a numerical value for each alternative and find the rank of alternative.

D. Assigning the six sigma software projects

Six sigma software projects are assigned to the relevant champions, including six sigma black belts and six sigma green belts. Each software project is allocating Six sigma to relevant champions, according to its very important step in the six sigma software projects. Consequently, in this step, it is interesting to assign software projects to qualified champions by taking into account the amount of software developers' effort and impact of software projects that are allowed in order to postpone some software projects. The help of fuzzy expert systems has extensively been maximized.



Fig.1 Triangular fuzzy logic number and the membership function



Table	1. Fuzzy logic parameters and	then triangular fuzz	ly logic numbers			
Fuzzy logic parameters fo	Fuzzy logic parameters for the ratings		c parameters	for	the	relative
		importance we	ghts of five cri	teria		
(0.000, 0.000, 1.000)	VP	(0.000, 0.000	, 1.000)	VL		
(0.000, 0.000, 1.000)	Р	(0.000, 0.100	, 0.300)	Ĺ		
(1.000, 3.000, 5.000)	MP	(0.100, 0.300	, 0.500)	ML		
(3.000, 5.000, 7.000)	F	(0.300, 0.500	, 0.700)	М		
(5.000, 7.000, 9.000)	MG	(0.500, 0.700	, 0.900)	MH		
(7.000, 9.000, 10.000)	G	(0.700, 0.900	, 1.000)	H		
(9.000, 10.000, 10.000)	VG	(0.900, 1.000	, 1.000)	VH		

Table 1. Fuzzy logic parameters and their triangular fuzzy logic numbers

III. PROPOSED SUPPORTING STRUCTURE FOR ASSIGNING AND CHOOSING THE SIX SIGMA SOFTWARE PROJECTS

The main objective of this article is to provide a supporting structure for prioritizing and assigning

The six sigma software projects for executing in organizational units. The phases of the presented supporting structure are illustrated in the following figure 2.



IV. CASE STUDY

Wipro software organization is one of the top software projects suppliers in India. In the first step of the presented supporting structure, very important software criteria were considered in the organizational unit. In this

regard, we considered 50 software developers, software managers of software development units, Six Sigma Champions, including black belts, green belts, quality assurance, and quality control experts, then the response rate was 80%, the validity of the questionnaire, such as

resources, relationship with strategic objectives, project scope etc. The reliability of the questionnaire was computed using $\alpha = 0.754$ d.f statistical hypothesis H0: $\mu i <= 3$,

H1: $\mu i>3$ is used to choose the software criteria with above-average importance. The t-test was done at the 0.05

Software criteria

t-test value

of significance 2-tailed L U Software 3 0.811 32 0.463 --0.2669 0.5678 resources availability 3 0.022 Software -2.199 32 -0.6842 -0.0433 investment return 0.599 0.550 -0.3111 0.5352 Special requirements to 3 32 improve 3 Software developers -2.515 32 0.016 -0.7225 -0.0862 satisfaction from software projects Time requirement 3 4.955 32 0.000 0.4911 1.1663 Scope of software 3 -4.699 32 0.000 -0.9663 -0.3662 projects Profits 3 0.211 32 0.850 -0.2938 0.3336 3 Relations with strategic 6.218 32 0.000 0.5776 1.1206 objectives -2.940 0.005 -0.7211 -0.1388 Measurability 3 32 Software customer 3 -2.793 32 0.010 -0.7982 -0.1316 satisfaction Key processes of the 3 0.454 32 0.650 -0.3100 0.6111 organization -0.0911 -2.600 0.013 -0.7672 Ability to access 3 32 information

Table 2. t-test results with 95% confidence level of signific	cance
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n

In the above t-test table2 shows the values of statistical hypothesis in those values of significance is more than 0.05, H0 is not rejected. In the above, for the six software criteria, those are more important than other criteria in this organizational unit. These software criteria will be used for

prioritizing the six sigma software project by the fuzzy logic TOPSIS technique.

level of significance, and n-1 d.f is investigated. The

95% confidence level

results are shown in the following table 2.

Significance

A. Choosing the six sigma potential software projects

In this section, for choosing the six sigma software potential projects, the software projects initial study is discussed in the studied software industry. Software potential projects, which have been removed from assessment reports of software customer satisfaction, software developer suggestion software competitor analysis, the software waste costs and the outcome of cost of quality, considered by software quality control and software quality assurance personnel of the software industry.

B. Prioritizing the six sigma software projects using fuzzy logic TOPSIS

Six potential software projects (p1,p2,p3,p4,p5, and p6) have been chosen for prioritization. These software projects have strengths and weaknesses. For prioritizing the six sigma software projects, six criteria's are:

- C1: Relation with the strategic objectives
- C2: Key processes of the organization
- C3: Profits
- C4: Special requirements to improve
- C5: Time requirement
- C6: Software resource availability

Among these criteria, C1, C2, C3, C4, and C6 are the most profited criteria, and C5 is the cost criteria. The advantage criterion means that a higher value is better, while, for the cost criterion, the reverse is valid. The executive software committee of three decision-makers, D1, D2, and D3, these decision-makers are experts and experienced in S/W quality control, and six sigma software projects are used to assess the importance of the software criteria, as shown in the following table 3.

Software criteria	Decision-makers			Mean weights
	D1	D2	D3	
C1	MP	MP	М	(0.030, 0.170, 0.370)
C2	G	G	G	(0.500, 0.700, 0.900)
C3	VG	VG	G	(0.630, 0.830, 0.970)
C4	Р	MP	Р	(0.009, 0.030, 0.170)
C5	VG	MG	G	(0.500, 0.700, 0.870)
C6	Н	Н	VG	(0.570, 0.770, 0.930)

Table 4. Fuzzy logic decision-making matrix and their fuzzy logic weights

	C1	C2	C3	C4	C5	C6
LV	(0.090,0.17	(0.500,0.700,	(0.630,0.830,0.9	(0.000,0.030,	(0.500,0.700,	(0.570,0.770
	0, 0.370)	0.900)	70)	0.170)	0.870)	,0.910)
P1	(0.330,1.67	(4.330,6.330,	(7.670,9.330,10.	(3.000,5.000,	(0.380,1.670,	(4.330,6.330
	0, 3.670)	8.330)	000)	7.000)	3.670)	,8.330)
P2	(1.670,3.67	(6.330,8.330,	(5.670,7.670,9.3	(0.000,0.330,	(1.670,3.670,	(5.670,7.670
	0,50670)	9.670)	30)	1.670)	5.670)	,9.330)
P3	(9.000,10.0	(3.670,5.670,	(5.670,7.670,9.3	(4.330,6.330,	(2.330,4.330,	(3.000,5.000
	00,10.000)	7.670)	30)	8.330)	6.330)	,7.000)
P4	(0.000,1.	(3.670,5.6	(9.000,10.000,	(5.000,7.0	(2.330,4.3	(0.000,0.6
	000,3.000)	70,7.670)	10.000)	00,9.000)	30,6.330)	70,2.330)

P5	(2.330,4.	(0.330,1.6	(8.330,9.670,1	(6.380,8.3	(6.330,8.3	(0.670,2.3
	330,6.330)	70,3.670)	0.000)	30,9.670)	30,9.670)	30,4.330)
P6	(6.330,8.	(0.000,0.0	(8.330,9.670,1	(4.330,6.3	(1.670,3.6	(8.330,9.6
	330,9.670)	00,1.000)	0.000)	30,8.330)	75,5.670)	70,10.000)

Table 5. weighted normalized fuzzy logic decision matrix

	C1	C2	C3	C4	C5 ⁻	C6
D1	(0.000.0	(0.220.0.460	(0.400.0.780.0.0	(0,000,0,020	(0.050.0.140	(0.250.0.490
r I	(0.000,0.	(0.220, 0.400, 0.780)	(0.490,0.780,0.9	(0.000,0.020,	(0.030,0.140,	(0.230,0.490,
	030,0.130)	0.780)	70)	0.120)	0.080)	0.780)
P2	(0.010,0.	(0.330,0.600,	(0.360,0.600,	(0.000,0.000,	(0.030,0.060,	(0.320,0.590,0.8
	060,0.210)	0.900)	0.900)	0.030)	0.050)	70)
P3	(0.030,0.1	(0.190,0.410,0.7	(0.360,0.640,0.9	(0.000,0.020,0.1	(0.030,0.050,0.0	(0.170,0.380,0.6
	70,0.370)	10)	00)	40)	50)	50)
P4	(0.000,0.0	(0.190,0.410,0.7	(0.570,0.830,0.9	(0.000,0.020,0.1	(0.030,0.050,0.0	(0.007,0.050,0.2
	20,0.110)	10)	70)	60)	50)	20)
P5	(0.010,0.0	(0.020,0.120,0.3	(0.530,0.810,0.9	(0.000,0.030,0.1	(0.020,0.030,0.0	(0.040,0.180,0.4
	70,0.230)	40)	70)	70)	30)	00)
P6	(0.020,0.1	(0.000,0.000,0.0	(0.530,0.810,0.9	(0.000,0.020,0.1	(0.030,0.060,0.0	(0.470,0.740,0.9
	40,0.350)	90)	70)	40)	50)	30)

Table 6. Positive and negative ideal solutions

	C1	C2	C3	C4	C5 ⁻	C6
\mathbf{P}^+	(0.370,0	(0.900,0.90	(0.970,0.97	(0.170,0.17	(0.140,0.14	(0.930,0.93
	.370,0.370	0,0.900)	0,0.970)	0,0.170)	0,0.140)	0,0.930)
)					
Р.	(0.000,0	(0.020,0.02	(0.360,0.36	(0.000,0.00	(0.020,0.02	(0.000,0.00
	.000,0.000	0,0.20)	0,0.360)	0,0.000)	0,0.020)	0,0.000)
)					

	Table 7. The distance of the altern	natives from the positive-nega	tive ideals and their final	weights and ranks
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Alternatives	\mathbf{D}^+	D.	$\mathbf{D}^+ + \mathbf{D}^-$	Final	Rank
				weight	
P1	1.7700	1.7470	3.5140	0.4970	2
P2	1.7200	1.8160	3.5350	0.5140	1
P3	1.9300	1.6330	3.5630	0.4580	3
P4	2.1500	1.2600	3.4140	0.3690	5
P5	2.2800	1.1590	3.4430	0.3370	6
P6	1.8900	1.5780	3.4670	0.4550	4

Table 8. Decision making normalized matrix

0.0.420.0.620 (0 170 0 270 0 570)	(0, 020, 0, 170, 0, 270)	(0.770, 0.020, 1.000)
0,0.430,0.030) (0.170,0.370,0.370)	(0.030, 0.170, 0.370)	(0.770, 0.930, 1.000)
0.0.930.1.000)	0.070.0.230.0.430)	(0.170.0.370.0.570)	(0.630.0.830.0.970)
(((0.0000,00000,0000,000)
0.0.020.1.000	0.200.0.500.0.700	(0.000.0.400.0.600)	(0.000.0.020.0.170)
0,0.930,1.000) (0.300,0.500,0.700)	(0.230, 0.430, 0.630)	(0.000, 0.030, 0.170)
(0, 0, 230, 0, 430) (0	0 770 0 930 1 000)	(0, 230, 0, 430, 0, 630)	(0.170, 0.370, 0.570)
(0.770,0.950,1.000)	(0.230,0.130,0.030)	(0.170,0.570,0.570)
0,0.030,0.170) (0.570,0.770,0.930)	(0.630, 0.830, 0.970)	(0.000, 0.030, 0.170)
0.04300630 (0 000 0 030 0 170)	(0.170.0.310.0.570)	(0, 300, 0, 500, 0, 700)
0,0.430,0.030)	0.000,0.030,0.170)	(0.170,0.510,0.570)	(0.500,0.500,0.700)
	0,0.430,0.630) (0,0.930,1.000) (0,0.930,1.000) (0,0.230,0.430) (0,0.030,0.170) (0,0.430,0.630) (0,0.430,0.630) (0.170,0.370,0.570) 0,0.930,1.000) (0.070,0.230,0.430) 0,0.930,1.000) (0.300,0.500,0.700) 0,0.230,0.430) (0.770,0.930,1.000) 0,0.030,0.170) (0.570,0.770,0.930) 0,0.430,0.630) (0.000,0.030,0.170)	0,0.430,0.630) (0.170,0.370,0.570) (0.030,0.170,0.370) 0,0.930,1.000) (0.070,0.230,0.430) (0.170,0.370,0.570) 0,0.930,1.000) (0.300,0.500,0.700) (0.230,0.430,0.630) 0,0.230,0.430) (0.770,0.930,1.000) (0.230,0.430,0.630) 0,0.030,0.170) (0.570,0.770,0.930) (0.630,0.830,0.970) 0,0.430,0.630) (0.000,0.030,0.170) (0.170,0.310,0.570)

Table 9. The value of impact and effort for each alternative

Projects	e	Effort Index	Impact Index	Index ES	Project allocation result
P1	(0.300,0.480,0.640)	0.47360	0.4970	0.4960	Green Belt
P2	(0.410,0.590,0.740)	0.5850	0.5140	0.6060	Black Belt
P3	(0.330,0.480,0.630)	0.4750	0.4580	0.4480	Green Belt
P4	(0.310,0.490,0.660)	0.4889	0.3690	0.3360	Give Up or Postpone
Р5	(0.300,0.420,0.560)	0.4208	0.3370	0.2960	Give Up or Postpone
P6	(0.180,0.330,0.520)	0.3375	0.4550	0.4440	Green Belt

V. SUMMARY AND CONCLUSION

The six sigma methodology is considered a software quality enhancement approach, including software waste reduction and profit-maximizing. The most important phases in executing a six sigma software project, the main purpose of this article are to develop a new model for choosing the most effective software projects in the industry and assigning them to software projects champions. The main purpose of this article is to develop a new model for choosing the most effective software projects in the industry and assigning them to software projects champions. In this article, using a line by line model, all the software criteria for six sigma software project allocation are considered, and the very important software projects are identified. Then, the

impact index, which is the outcome of the TOPSIS technique, and the effort index are computed.

Using the fuzzy logic expert system, the six sigma software projects are allocated to the relevant experts or champions. It can be claimed that two indexes, such as impact and also effort, are used for software projects allocation and the six sigma software potential projects to six sigma champions. To compute the impact index, the important software criteria such as resource availability, time requirement, strategic objectives relations, key processes relation in software organizations, profits, and special requirements to improve. In the case study, some of the objectives are very important. They are shown in table 2.

In six sigma software project selection and prioritization, some of the objectives are very important such as the direct relation with strategic objectives, profits, and time requirements. The effort index and the impact index are effective on six sigma software project allocation and prioritization.

Further work can concentrate on the evaluation of alpha and beta in the fuzzy logic expert system. Supporting structure can be used as a fundamental pattern for developing new models in other organizations. And very important criteria in the allocation and prioritization of six sigma projects can be mentioned in these organizations.

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Chandrakanth G Pujari / IJCTT, 68(2), 25-33, 2020