Material Selection of Machine Design using Expert System: A Comparative Study

Prithwiraj Jana



Abstract: This research venture one of the major concerns in the field of expert system. Material selection an important key issue of machine design. Objectives of computerized selection procedure are reduced to personal bias and gives the more accurate optimized result. The concept of entropy; to evaluate the weight factor for each alternative material property or performance index, and the other is TOPSIS and SAW; to rank the candidate materials, for which several requirements are considered simultaneously. Sensitivity analysis is introduced here for better performance of selection.

Keywords: Flywheel, Material selection, TOPSIS, SAW, Entropy, Sensitivity analysis

I. INTRODUCTION

Over the years, selection of materials plays a crucial role for value addition in production, cost optimization, better product development and improve productivity in machine design. Flywheel is an important machining component. In this paper, optimization theory is applied for right selection of flywheel manufacturing process. Multiple criteria decision making (MCDM) is the process of selecting the best alternative from a set of feasible alternatives considering multiple conflicting criteria. In precise terms criteria are considered to be 'strictly' conflicting if the increase in satisfaction of one result in a decrease in satisfaction of the other.

1.1 Entropy

Originally Entropy is a thermodynamic concept, first introduced by Shannon. It has been widely used in the engineering, socioeconomic and other fields. According to the basic principles of information theory, information is a measure of system's ordered degree, and the entropy is a measure of system's disorder degree. Step1Calculate pij (the ith scheme's jth indicator value's proportion). pij =rij / $\sum_{j=1}^{m}$ = rij, rijis the ith scheme's jth indicator valueStep2 Calculate the jth indicator's entropy value becomesej.ej= -k $\sum_{j=1}^{m}$ pijln pij, k=1/ln m, m is the number of assessment schemes.Step3 Calculate weight wj (jth indicator's weight).

wj=(1-ej) $\sum_{j=1}^{n} (1 - e_j)$, n is the number of indicators, and $0 \le w_j \le 1, \sum_{j=1}^{n} w_j = 1$

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In entropy method, the smaller the indicator's entropy value ej is, the bigger the variation extent of assessment value of indicators is, the more the amount of information provided, the greater the role of the indicator in the comprehensive evaluation, the higher its weight should be.

1.2 Technique for Order Preference by Similarity to Ideal Solution (TOPSIS)

TOPSIS is an evaluation method that is often used to solve MCDM problems. It has a number of applications in practice, such as comparison of company performances, financial ratio performance within a specific industry and financial investment in advanced manufacturing systems, etc. However, there are also some limits to it. So far, the work on how to improve original TOPSIS method has mainly emphasized on improving the weight to sensitize the R value. Besides, there has also been improvement on formula of the R value, such as the 'Miqiezhi' method. Because of the complexity of evaluation problems, a better and simpler method is required to understand the inherent relationship between the R value and alternative evaluation. In this report, a novel, modified TOPSIS (M-TOPSIS) method is described as a process of calculating the distance between the alternatives and the reference points in the D+ D--plane and constructing the R value to evaluate quality of alternative.

1.3 SIMPLE ADDITIVE WEIGHTING (SAW)

Step 1 Formation of decision matrix: Criterion outcomes of decision alternatives can be collected in a table called Decision Matrix comprised of a set of columns and rows. The matrix rows represent decision alternatives, with matrix columns representing criteria. A value found at the intersection of row and column in the matrix represents a criterion outcome - a measured or predicted performance of a decision alternative on a criterion. The decision matrix is a central structure of the MCDA/MCDM since it contains the data for comparison of decision alternatives.

1.4 Sensitivity Analysis

In actual situation decision-making is rather dynamic process not static. It varies in the continuous changing environment. In reality the value of decision making attitude depends upon decision maker's personal choice. Under such circumstances decision making attitude behaves as a variable that may yield different results. Keeping it in mind, the proposed model for the selection of material has been enhanced by sensitivity analysis to provide a readymade solution of the current problem under variable decision making attitude. The governing equation of the material measure (MM) is given by $MM = \alpha(OEM - SEM) + SEM$

 $MM_i = \alpha (OFM_i - SFM_i) + SFM_i$



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where, i = 1, 2...m.

OFMi= Objective factor measure for the alternative i SFMi= Subjective factor measure for the alternative i α = Objective factor decision weight/Coefficient of attitude

II. PROCEDURE OF CASE STUDY

An organization has got 10 different materials with different specifications for flywheel. The decision maker considered 5 selection criteria.

The materials are as follows

The selection criteria are as follows:

C1Density (g/cm3)C2Hardness (BHN)C3Young's modulus (GPa)C4Bulk Modulus (GPa)C5Poisson's ratioC6Cost (INR) Per kg	Table-01			
C2 Hardness (BHN) C3 Young's modulus (GPa) C4 Bulk Modulus (GPa) C5 Poisson's ratio C6 Cost (INR) Per kg	C1	Density (g/cm3)		
C3 Young's modulus (GPa) C4 Bulk Modulus (GPa) C5 Poisson's ratio C6 Cost (INR) Per kg	C2	Hardness (BHN)		
C4 Bulk Modulus (GPa) C5 Poisson's ratio C6 Cost (INR) Per kg	C3	Young's modulus (GPa)		
C5 Poisson's ratio C6 Cost (INR) Per kg	C4	Bulk Modulus (GPa)		
C6 Cost (INR) Per kg	C5	Poisson's ratio		
	C6	Cost (INR) Per kg		

Out of 6 criteria, 4 criteria viz. C2:Hardness (Bhn), C3: Young's modulus (GPa),C4: Bulk Modulus (GPa),C5: Poisson's ratio are beneficial criteria because their higher values are desirable and remaining viz. C1: Density (g/cm3) C7: Cost (INR) Per kg are non-beneficial criteria because their lower values are desirable.

The objective of the decision maker is to assess the performance of the materials. Counseling the above 6 criteria

			Table-2		
MATERIAL	σlimit (MPa)	KIC (MPa/m1/2)	Density 103 kg/m3	Price/mass (US\$/Ton)	Fragmentability
300M	800	68.9	8	4200	3
2024-T3	140	38	2.82	2100	3
7050-T73651	220	35.4	2.82	2100	3
Ti-6Al-4V	515	123	4.73	10500	3
E glass]epoxy FRP	140	20	2	2735	9
S glass]epoxy FRP	330	50	2	4095	9
Carbon]epoxy FRP	700	35	1.59	35470	7
Kevlar 29-epoxy FRP	340	40	1.4	11000	7
Kevlar 49-epoxy FRP	900	50	1.46	25000	7
Boron]epoxy FRP	1000	46	2	315000	5
. 1 1 1	1 1	1.1 771. 1.1.1.1.	1		

to ultimately select the best material. The decision maker applied TOPSIS and MOORA methods for their simplicity, adaptability, applicability and is of applications. The decision matrix for the materials with respect to the criteria shown below:

Computational result by MATLAB:

2.1 Entropy Method:

III. RESULT

					Table-3					
	ENTROPY METHOD									
crit	eria		C1	C2	Ca	3		C4	C5	
weighte	d values		0.2729	0.1019	- 0.095	1	0.31	152	0.4051	
2.2 SAW N	IETHOD):			1.0000	0.3740	0.2500	0.0067	0.6000	
RESULT:				STEP 2:	Determ	ination of	weighted	normalized decision	l	
STEP1: 1	Determin	ation of 1	normalize	d decision matrix	matrix			Ū.		
0.8000	0.5602	1.0000	0.5000	1.0000	0.2183	0.0571	-0.0951	0.1576	0.4051	
0.1400	0.3089	0.3525	1.0000	1.0000	0.0382	0.0315	-0.0335	0.3152	0.4051	
0.2200	0.2878	0.3525	1.0000	1.0000	0.0600	0.0293	-0.0335	0.3152	0.4051	
0.5150	1.0000	0.5913	0.2000	1.0000						
0.1400	0.1626	0.2500	0.7678	0.3333						
0.3300	0.4065	0.2500	0.5128	0.3333					\frown	
0.7000	0.2846	0.1988	0.0592	0.4286					nd Neural Netw	
0.3400	0.3252	0.1750	0.1909	0.4286					Se a A	
0.9000	0.4065	0.1825	0.0840	0.4286				Intellige	IJAINN	

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0.1406	0.1019	-0.0563	0.0630	0.4051	
0.0382	0.0166	-0.0238	0.2420	0.1350	
0.0901	0.0414	-0.0238	0.1617	0.1350	
0.1910	0.0290	-0.0189	0.0187	0.1736	
0.0928	0.0331	-0.0166	0.0602	0.1736	
0.2456	0.0414	-0.0174	0.0265	0.17360.2729	
0.0381 -	0.0238	0.0021 ().2431		
STEP 3:	Compu	tation of c	omposite	score sby sum o	of
all weight	ed norma	lized row	s		
TT1 1	6()				

The values of (s) are:



Fig:01

2.3 To determine the sensitivity analysis graph by SAW: The value of closeness co-efficient in SAW method when alpha=0 when alpha=1

1	
0.5627	0.1803
0.7203	0.0361
0.7203	0.0558
0.4681	0.1862
0.3771	0.0310
0.2967	0.1077
0.1923	0.2011
0.2338	0.1093
0.2001	0.2697
0.2452	0.2872



Fig:02

2.4 TOPSIS method In the TOPSIS method

The weighted values are: 0.2729 0.1019 -0.0951 0.3152 0.4051
1.2 The weighted values got from entropy method STEP1: Determination of normalized decision

matrix

0.8000	0.5602	1.0000	0.5000	1.0000
0.1400	0.3089	0.3525	1.0000	1.0000
0.2200	0.2878	0.3525	1.0000	1.0000
0.5150	1.0000	0.5913	0.2000	1.0000
0.1400	0.1626	0.2500	0.7678	0.3333
0.3300	0.4065	0.2500	0.5128	0.3333
0.7000	0.2846	0.1988	0.0592	0.4286
0.3400	0.3252	0.1750	0.1909	0.4286
0.9000	0.4065	0.1825	0.0840	0.4286
1.0000	0.3740	0.2500	0.0067	0.6000
STEP 2:				

Determination of positive ideal solution: taking the maximum values of each column from the normalized decision matrix

1 1 1 1 1

Determination of negetive ideal solution: taking the minimum values of each column from the normalized decision matrix

0.1400 0.1626 0.1750 0.0067 0.3333 STEP 3:

Calculation of the seperation measure from the positive ideal solution(di_Plus)

0.3308,,0.4589,,0.4217,,0.5001,,0.6456,,0.5998,,0.6 534,,0.6627,,0.6098,,0.6019

Calculation of the seperation measure from the negetive ideal solution(di_Minus)

0.5718,, 0. 7002,, 0.7010,, 0.5340,, 0.4267,,

0.3101,, 0.3026,, 0.1673,, 0.4114,, 0.4844

Calculation of R_i



Fig:03

2.5 Comparative analysis of ranking of flywheel materials using MCDM methods



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	Table-4	
MATERIAL	MOORA	TOPSIS
	(RANK)	(RANK)
M1	2	1
M2	3	3
M3	5	5
M4	1	2
M5	4	4

IV. CONCLUSION

The use of TOPSIS and MOORA methods are inspected to be quite capable and computationally easy to appraise and select the proper material from a given set of alternatives. These methods use the measures of the considered criteria with their relative importance in order to arrive at the final ranking of the alternative flywheel Materials. We have used the MATLAB, by this software we can also make rank of any system for any number of alternatives and criteria within a fraction of second with accuracy. Thus, these popular MCDM methods can be successfully employed for solving any type of decision-making problems having any number of criteria and alternatives in the manufacturing domain. In this study, cost is an important key issue in material selection. For a product development, cost optimization is necessary.

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