

# Research on Strategic Supplier Selection and Evaluation Method of Manufacturing Enterprise Based on AHP-GCA Method

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**Abstract.** With the application of intelligence technology in manufacturing industry, the competition of supply chain in manufacturing industry is becoming more and more competitive. Strategic supplier is a key element for the success of manufacturing enterprise. At present, there are many methods for the selection and evaluation of strategic supplier of manufacturing enterprise. Analytical hierarchy process and grey correlation analysis (AHP-GCA) are methods used to choose strategic supplier of manufacturing enterprise by building optimum reference data and solving grey correlation degree. These have positive effects to the enterprise practice in qualitative diagnosis, quantitative diagnosis and grey information processing.

## Introduction

The standards of strategic supplier selection and evaluation system have obvious characteristic of hierarchy. The first level standards can be disassembled into different sub-indexes which form tree structure that offers structure basis for analytic hierarchy method. Analytic hierarchy method is a relatively mature theory which has much hands-on experience to draw on. Grey correlation analysis is a method which carries out quantitative description and comparison towards system development and change. Its basic idea is to judge whether the relation between reference sequence of number and comparison sequence of number is close or not by confirming the degree of closeness/ proximity of these two curves, and use grey correlation degree to reflect the proximity between curves and find out the difference and closeness. AHP-GCA (Analytic hierarchy processing and grey correlation analysis) model combine analytic hierarchy processing method with grey correlation analysis method organically. In AHP-GCA (Analytic hierarchy processing and grey correlation analysis) model, different indexes' weight are confirmed by using analytic hierarchy processing method. Human's subjective judgment is expressed in number, and consistency check is carried out, thus the side-effect of subjective factors could be brought down, and drastically. Grey correlation analysis method implements whitening processing on qualitative index items (grey indexes) from indexes, and standardizes quantitative index data. By this mean, the losing and deviation of information would be minimized and the correctness of conclusion could be guaranteed.

## Selecting Evaluation Model Based on AHP-GCA

First step, building tree structure model.

Second step, constructing "multiple judgment" matrix,  $B_k$  signifies number  $k$  index in the  $B$  gradation. Suppose  $k$ 's indexes at this gradation is  $B_1, \dots, B_n$ , its direct upper level index is  $A$ , then make multiple comparison to  $B_1, \dots, B_n$  aiming at index  $A$  by using nine points marking method. Thus comparison value  $b_{ij}$  is appeared. We note down judgment matrix  $B = (b_{ij})_{n \times n}$ . So  $B$ 's maximum eigenvalue is  $\lambda_{\max}$ , normalized eigenvector that belongs to  $\lambda_{\max}$  is  $\omega = (\omega_1, \dots, \omega_n)^T$ . In this way,  $(\omega_1, \dots, \omega_n)$  is weighting of index  $B_1, \dots, B_n$  to index  $A$ . In actual computation, we can calculate

approximately index relative weighting under every single index by using “extraction of root” method. And after normalization, it is relatively important weighting of indexes at the same gradation to indexes at the direct upper gradation. In the similar way, we calculate from upper gradation to lower gradation until it comes to the last gradation which has its weighting to the direct upper gradation. If we note  $D_k$  as matrix organized by weighting in column of all indexes in number k gradation to all indexes in the direct upper gradation, then the combined weighting vector in gradation is:

$$W^k = D_k \cdot D_{k-1} \cdot \dots \cdot D_2 \cdot D_1 . \quad (1)$$

While getting judgment matrix B, index consistency  $C.I$  is needed check due to probable existence of subjective judgment inconsistency of nine points marking method.

In it,

$$\overline{\omega}_i = \sqrt[n]{\prod_{j=1}^n b_{ij}} . \quad (2)$$

$$\omega_i = \frac{\overline{\omega}_i}{\sum_{i=1}^n \overline{\omega}_i} . \quad (3)$$

$$C.I = \frac{\lambda_{\max} - n}{n - 1} . \quad (4)$$

$$\lambda_{\max} = \sum_{i=1}^n \frac{\sum_{j=1}^n b_{ij} \omega_j}{n \omega_i} . \quad (5)$$

If  $C.I \leq 0.1$ , then we judge matrix B is satisfactory[1].

The third step, collecting various evaluation indexes data of back up strategic suppliers, and carrying out standardization on quantitative indexes data, quantitative processing of whitening weight function on qualitative indexes data. After this, the indexes data sequences constitute comparison data sequence of evaluation index of various back up suppliers. Choose the maximum values (optimum values) from various indexes that have been processed to form the optimal reference data  $\{ Z_0 \} = \{ z_{01}, z_{02}, \dots, z_{0n} \}$ .

(1) Standardization of quantitative index data

Extremely large index --the larger the indexes value the better

Extremely small index--the smaller the indexes value the better

Suppose there are g back up suppliers, f indexes.  $d_{ij}$  is original data,  $y_{ij}$  is standard data ( $i=1, 2, \dots, g$ ;  $j=1, 2, \dots, f$ ).  $G_j$  and  $g_j$  are respectively the largest value and the smallest value of number j index of back up supplier. The indexes data is changed refers with Fig.1 and Fig. 2 .

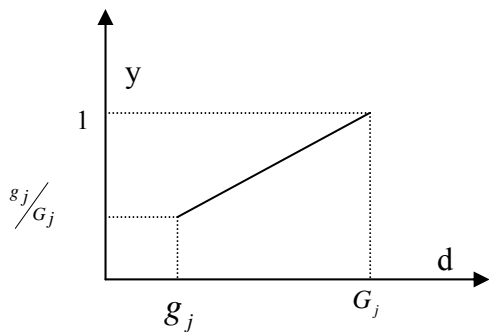


Fig.1 Extremely large type

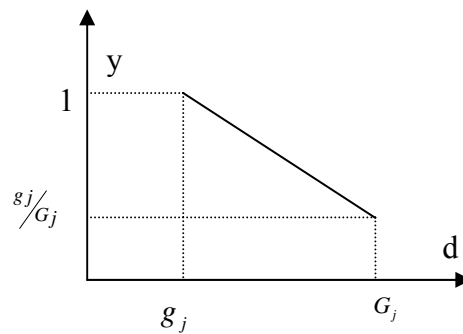


Fig.2 Extremely small type

Extremely large index data conversion:

$$y_{ij} = d_{ij} / G_j \quad (i=1,2,3,\dots,g; j \in \text{extremely large type}). \quad (6)$$

Extremely small index data conversion:

$$y_{ij} = 1 + g_j / G_j - y_{ij} / G_j \quad (i=1,2,3,\dots,g; j \in \text{extremely small type}). \quad (7)$$

## (2) Quantitative processing of whitening weight function on qualitative indexes data

Experts make scores at one qualitative index respectively according to the above regulated evaluation standards. Making lowest score and the highest score value 0 and 1(or vice versa) is to solve the whitening value of such index. The whitening value are solved according to Fig. 3 and Fig. 4 showed below.

Suppose there are  $g$  backup suppliers, qualitative index  $A$ ,  $H$  experts make scores on it respectively according to the evaluation standards.

If index  $A$  is extremely large index-- the larger the indexes value the better

When  $y = y_{\max}$ , whitening value  $x = 1$

When  $y = y_{\min}$ , whitening value  $x = 0$

If index  $A$  is extremely small index-- the smaller the indexes value the better, then

when  $y = y_{\max}$ , whitening value  $x = 0$

when  $y = y_{\min}$ , whitening value  $x = 1$

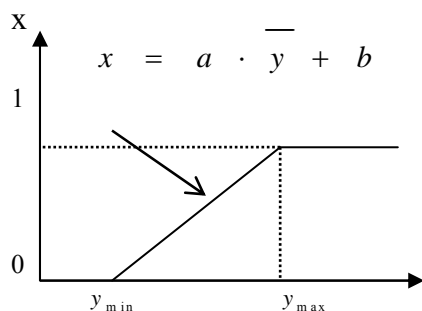


Fig.3 Extremely large type

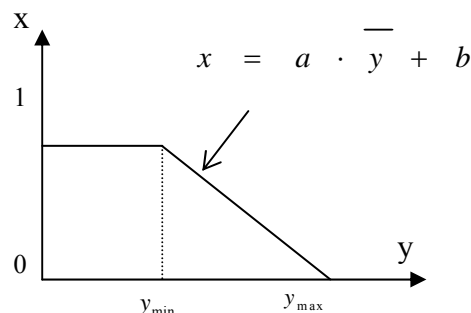


Fig.4 Extremely large type

The fourth step, calculating grey correlation coefficient and grey correlation degree.

Correlation data  $\beta_{ij}$  is difference value in the number  $j$  index of  $\{ Z_i \}$  which is comparison sequence of number of backup supplier evaluation index to  $\{ Z_0 \}$  which is the optimal reference data. The difference value can be used as yard stick of correlation degree. The larger the  $\beta_{ij}$ , the more correlated the two sequence are in the number  $j$  index.

$$\beta_{ij} = \frac{\min_i \min_j \Delta_{ij} + \rho \max_i \max_j \Delta_{ij}}{\Delta_{ij} + \rho \max_i \max_j \Delta_{ij}}. \quad (8)$$

In the formula  $\Delta_{ij} = |z_{0j} - z_{ij}|$ ,  $\rho$  is called Distinguishing Coefficient. It is for weakening the effects of distortion due to over large of the largest absolute difference. The smaller the value of  $\rho$  is, the greater the distinguishing capability. It is chosen between 0 and 1, usually  $\rho = 0.5$ . the chosen value of  $\rho$  does not change correlation degree [2].

Because correlation coefficient can only reflect correlation degree of comparison data column with the optimal reference data in one selection and evaluation index, it can not embody the advantage and disadvantage of backup suppliers. So we can use correlation degree to embody it, correlation degree between comparison data column  $\{ Z_i \}$  and the optimal reference data  $\{ Z_0 \}$  is

$$\theta_i = 1 / n \sum_{i=1}^n \omega_i \beta_{ij}. \quad (9)$$

Fifth step, calculating grey correlation degree of backup suppliers according to the above steps and getting the results in order. The larger the grey correlation degree value  $\theta_i$  is the more close comparison data column  $\{ Z_i \}$  and the optimal reference data  $\{ Z_0 \}$  is in space shape. That is, the more powerful the backup supplier is in overall strength. Since it could be chosen as important partner model for strategic supplier for manufacturing enterprise.

## Sample

Located in Anhui province, C Automobile Science and Technology Corporation Ltd which integrated vehicle designing and manufacturing is comprehensive business corporation established in 2003 and invested by both of one Chinese finished automobile enterprise and one Canadian automobile parts enterprise. It is in the leading place in the field of vehicle power-operated sliding door, and provides its products to one MPV vehicle manufacturer. The corporation applied the above mentioned strategic supplier selection and evaluation system of manufacturing enterprise in the work of selection and evaluation of power-operated sliding door backup supplier and did a nice job.

C corporation chose staff from department of purchasing, research and development, and production to organize evaluation team for sliding door drive system supplier selection. Then they decided 5 suppliers distributed Yangtze River delta as the backup suppliers after a round of primary election. Drawing lessons from the suppliers relationship management mode of General Motors, C corporation carried out on-site investigation of one week to collect related information and use AHP-GCA (analytic hierarchy process-grey correlation analysis) to carry out selection and evaluation to achieve the advantages and disadvantages of these suppliers.

First, constructing multiple judgment matrix, scoring by experts through nine points ratio method, calculating weighting of various indexes and carrying out consistency check. Table 1 shows that building 7 first level indexes and 23 second level indexes based on selection and evaluation index system. Various indexes and weighting at Index level B and level C and consistency check results are showed respectively from Table 2 to Table 9.

Table 1. C level indexes 's combined weighting towards target A

	B level indexes weighting		C level indexes weighting		C C level indexes 's combined weighting
Supplier strategic Selection and evaluation index system of C corporation power-operated sliding door system	Product competitive edgeB1		Quality system C1	0.2803213	0.083091429
			Product failure rate C2	0.1441862	0.042738958
		0.296415	synthetic cost C3	0.2079525	0.061640244
			Delivery on time rateC4	0.1542667	0.045726975
			Production capabilityC5	0.0908316	0.026923857
			Market share C6	0.1224416	0.036293525
	Management level B2		Cost control level C7	0.4599581	0.086585543
		0.1882466	Inventory control level C8	0.2211248	0.041625988
			Service level C9	0.3189171	0.060035063
	Flexibility evaluation B3	0.1276272	volume of production flexibility C10	0.25	0.031906809
			Delivery flexibility C11	0.75	0.095720428
	Research and development ability B4		Scientific research expense rate C12	0.1634241	0.01487299
		0.091009	New product development success rate C13	0.2969613	0.027026016
			New product sale rateC14	0.5396146	0.04910953
	Manpower B5		Professional title composition C15	0.25	0.017613987
		0.070456	Education background composition C16	0.25	0.017613987
			Training fee ration C17	0.5	0.035227974
	Financial condition B6		Current ratioC18	0.5396146	0.04910953
		0.091009	Return on total asset rate C19	0.1634241	0.01487299
			Ratio of liabilities to assets C20	0.2969613	0.027026016
	Enterprise compatibility B7		Development strategy compatibility C21	0.3108137	0.042033872
		0.135238	Enterprise culture compatibility C22	0.493386	0.066724612
			Information platform compatibility C23	0.1958004	0.02647968

Table 2. Various indexes and weighting at Index level B and consistency check

A	B1	B2	B3	B4	B5	B6	B7	$\omega_i$	$\lambda_i$
B1	1	2	3	3	3	2	3	0.296415	1.02554815
B2	1/2	1	1	1	3	3	3	0.1882466	1.09673607
B3	1/3	1	1	2	2	2	1/3	0.1276272	1.07982352
B4	1/3	1	1/2	1	1	1	1/2	0.0910085	1.05320939
B5	1/3	1/3	1/2	1	1	1/2	1/2	0.0704559	1.0137141
B6	1/2	1/3	1/2	1	1	1	1	0.0910085	1.03990428
B7	1/3	1/3	3	2	2	1	1	0.1352382	1.15522256
$\lambda_{\max}=7.46415806$				CI=0.07735968				1	

Table 3. Various indexes and weighting at Index level C and consistency check

B1	C1	C2	C3	C4	C5	C6	$\omega_i$	$\lambda_i$
C1	1	3	1	2	3	2	0.280321	1.038537
C2	1/3	1	1	1	2	1	0.144186	1.044888
C3	1	1	1	1	3	2	0.207953	0.7872638
C4	1/2	1	1	1	1	2	0.154267	1.0612364
C5	1/3	1/2	1/2	1	1	1/2	0.090832	1.0565867
C6	1/2	1	1/2	1/2	2	1	0.122442	1.0475217
$\lambda_{\max}=6.0360337$				CI=0.0072067			1	

Table 4. Various indexes and weighting at Index level C and consistency check

B2	C7	C8	C9	$\omega_i$	$\lambda_i$
C7	1	3	1	0.46	1.05
C8	1/3	1	1	0.221	1.05
C9	1	1	1	0.319	1.05
$\lambda_{\max}=3.14$		CI=0.07		1	

Table 5. various indexes and weighting at Index level C and consistency check

B3	C10	C11	$\omega_i$	$\lambda_i$
C10	1	1/3	0.25	1
C11	3	1	0.75	1
			1	
$\lambda_{\max}=2$		CI=0		

Table 6. Various indexes and weighting at Index level C and consistency check

B4	C12	C13	C14	$\omega_i$	$\lambda_i$
C12	1	1/2	1/3	0.163	1
C13	2	1	1/2	0.297	1
C14	3	2	1	0.54	1
$\lambda_{\max}=3.01$		CI=0		1	

Table 7. Various indexes and weighting at Index level C and consistency check

B5	C15	C16	C17	$\omega_i$	$\lambda_i$
C15	1	1	1/2	0.25	1
C16	1	1	1/2	0.25	1
C17	2	2	1	0.5	1
$\lambda_{\max}=3.01$		CI=0		1	

Table 8. Various indexes and weighting at Index level C and consistency check

B6	C18	C19	C20	$\omega_i$	$\lambda_i$
C18	1	3	2	0.54	1.0031
C19	1/3	1	1/2	0.163	1.0031
C20	1/2	2	1	0.297	1.0031
$\lambda_{\max}=3.0092$		CI=0.0046		1	

Table 9. Various indexes and weighting at Index level C and consistency check

B7	C21	C22	C23	$\omega_i$	$\lambda_i$
C21	1	1/2	2	0.311	1.0179
C22	2	1	2	0.493	1.0179
C23	1/2	1/2	1	0.196	1.0179
$\lambda_{\max}=3.0536$		CI=0.0268		1	

Second, carrying out standardization and whitening weight function on five backup suppliers, see table 10.

Table 10. Power-operated sliding door suppliers original data/comparison sequence of number

Suppliers	A		B		C		D		E		Z <sub>0</sub>
original data/comparison sequence of number	Original	Z <sub>1</sub>	Original	Z <sub>2</sub>	Original	Z <sub>3</sub>	Original	Z <sub>4</sub>	Original	Z <sub>5</sub>	
Quality system C1	9	1	8	0.5	9	1	9	1	8	0.5	1
Product b defect rate C2	300	1	350	0.88	400	0.75	300	1	350	0.88	1
Synthetic cost C3	3780	0.98	3880	0.95	3960	0.93	3690	1	3750	0.99	1
Delivery on time rate C4	98%	0.99	99%	1	99%	1	97%	0.98	98%	0.99	1
Production capacity C5	100	0.5	200	1	150	0.75	130	0.65	160	0.8	1
Market share C6	15%	0.83	18%	1	10%	0.56	8%	0.44	12%	0.67	1
Cost control level C7	130%	0.96	125%	0.93	135%	1	128%	0.95	120%	0.89	1
Inventory control level C8	86%	0.59	108%	0.74	133%	0.91	120%	0.82	147%	1	1
Service level C9	13%	0.62	10%	0.85	8%	1	9%	0.92	11%	0.77	1
Volume of production flexibility C10	80%	0.88	87%	0.96	91%	1	86%	0.95	78%	0.86	1
Delivery flexibility C11	28%	0.93	30%	1	22%	0.73	25%	0.83	23%	0.77	1
Scientific research expense rate C12	5%	1	3%	0.6	4%	0.8	2%	0.4	5%	1	1
New product development success rate C13	70%	0.92	68%	0.9	76%	1	59%	0.78	63%	0.83	1
New product sale ratio C14	10%	0.67	12%	0.8	8%	0.53	15%	1	11%	0.73	1
Professional title composition C15	38%	0.88	42%	0.98	40%	0.93	39%	0.91	43%	1	1
Education background composition C16	68%	0.93	71%	0.97	66%	0.9	68%	0.93	73%	1	1
Training fee ratio C17	0.20%	0.67	0.30%	1	0.25%	0.83	0.23%	0.77	0.19%	0.63	1
Liquidity ratio C18	2	0.89	2.1	0.93	1.95	0.87	2.2	0.98	2.25	1	1
Return on total asset rate C19	12%	1	11%	0.92	10%	0.83	11%	0.92	12%	1	1
Ratio of liabilities to assets C20	45%	0.88	48%	0.81	39%	1	42%	0.94	43%	0.92	1
Development strategy compatibility C21	7	0.5	8	1	7	0.5	8	1	7	0.5	1
Enterprise culture compatibility C22	8	1	7	0.5	7	0.5	8	1	7.5	0	1
Information platform compatibility C23	8	0.5	9	1	8	0.5	8	0.5	9	1	1

Third, solving grey correlation coefficient  $\beta_{ij}$  and grey correlation degree  $\theta_i$ , see Table 11.

Table 11. Power-operated sliding door suppliers  $\Delta_{ij}$ ,  $\beta_{ij}$  and  $\theta_i$  ( $\rho=0.5$ )

suppliers	A		B		C		D		E	
comparison sequence of number	$\beta_{i1}$	$\Delta_i(1)$	$\beta_{i2}$	$\Delta_i(2)$	$\beta_{i3}$	$\Delta_i(3)$	$\beta_{i4}$	$\Delta_i(4)$	$\beta_{i5}$	$\Delta_i(5)$
Quality system C1	1.000	0	0.357	0.5	1.000	0	1.000	0	0.357	0.5
Product b defect rate C2	1.000	0	0.690	0.13	0.527	0.25	1.000	0	0.690	0.13
Synthetic cost C3	0.924	0.02	0.853	0.05	0.803	0.07	1.000	0	0.949	0.02
Delivery on time rate C4	0.965	0.01	1.000	0	1.000	0	0.933	0.02	0.965	0.01
Production capacity C5	0.357	0.5	1.000	0	0.527	0.25	0.443	0.35	0.582	0.2
Market share C6	0.625	0.17	1.000	0	0.385	0.44	0.333	0.556	0.455	0.33
Cost control level C7	0.883	0.04	0.790	0.07	1.000	0	0.842	0.05	0.715	0.11
Inventory control level C8	0.401	0.42	0.512	0.27	0.745	0.1	0.602	0.18	1.000	0
Service level C9	0.419	0.39	0.644	0.15	1.000	0	0.783	0.08	0.546	0.23
Volume of production flexibility C10	0.697	0.12	0.863	0.04	1.000	0	0.835	0.06	0.660	0.14
Delivery flexibility C11	0.806	0.07	1.000	0	0.510	0.27	0.625	0.17	0.544	0.23
Scientific research expense rate C12	1.000	0	0.410	0.4	0.582	0.2	0.357	0.6	1.000	0
New product development success rate C13	0.779	0.08	0.726	0.11	1.000	0	0.554	0.22	0.619	0.17
New product sale ratio C14	0.455	0.33	0.582	0.2	0.373	0.47	1.000	0	0.510	0.27
Professional title composition C15	0.706	0.12	0.924	0.02	0.799	0.07	0.749	0.09	1.000	0
Education background composition C16	0.803	0.07	0.911	0.03	0.743	0.1	0.803	0.07	1.000	0
Training fee ratio C17	0.455	0.33	1.000	0	0.625	0.17	0.544	0.23	0.431	0.37
Liquidity ratio C18	0.715	0.11	0.806	0.07	0.676	0.13	0.927	0.02	1.000	0
Return on total asset rate C19	1.000	0	0.770	0.08	0.625	0.17	0.770	0.08	1.000	0
Ratio of liabilities to assets C20	0.690	0.13	0.598	0.19	1.000	0	0.818	0.06	0.770	0.08



Development strategy compatibility C21	0.357	0.5	1.000	0	0.357	0.5	1.000	0	0.357	0.5
Enterprise culture compatibility C22	1.000	0	0.357	0.5	0.357	0.5	1.000	0	0.357	0.5
Information platform compatibility C23	0.357	0.5	1.000	0	0.357	0.5	0.357	0.5	1.000	0
$\min \Delta_{ij}$		0		0				0		0
$\max \Delta_{ij}$		0.5		0.5		0.5		0.556		0.5
$\theta_i$	0.032		0.033		0.031		0.035		0.029	

Fourth, sequence according to grey correlation degree.

According to grey correlation degree  $\theta_i$  from large value to small value,  $\theta_4 > \theta_2 > \theta_1 > \theta_3 > \theta_5$ .

Among five backup power-operated sliding door drive system suppliers of C corporation, D supplier is benchmarking. Of course, D supplier with the largest correlation degree may not be the one which suits C corporation best. C corporation has to choose alternatives, B and A can be chosen as backup alternatives. In the actual work, C corporation finally chooses D supplier as its strategic supplier of power-operated sliding door.

## Conclusion

Strategic supplier selection is one of the important questions within supply chain in manufacturing industry. The integrated AHP-GCA method for supplier selection and evaluation is effective. First, using the method of AHP to calculate the weights of the first level indexes and the second level indexes and checking the consistency in order to combine the experience of industry experts with their rational thinking. Second, through standardization and whitening weight function of quantitative and qualitative data, the comparison data of optional suppliers are formed. Then using grey correlation method to solve correlation coefficient of various optional suppliers by combining the weights of all indexes. In this way, the order of advantage and disadvantage are confirmed. It makes great achievement in quantitative and qualitative decisions and grey information disposal in actual application of selection and evaluation of optional strategic supplier in manufacturing industries.

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