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Reduction and Optimization of Supplier Risk Indicators Based on Rough Set

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Abstract: The determination of supplier risk indicators is complex. Using vast data from SAP system of the enterprise, risk warning indicators can be reduced and optimized by the method of rough set. First of all, extract historical data from SAP system, and determine the discrete rules as excellent, good, moderate, and poor for each risk indicators to construct knowledge set which can be used for rough set operation. Then, using rough set theory to divide decision attribute set into equivalence classes, reduce non essential attributes, and calculate the dependence and importance degree for each essential attributes. After the normalization for all essential attributes, the reduced and optimized indicators for supplier risk evaluation system can be reached.

Keywords: Attribute reduction, discretization, optimization, rough set, supplier risk.

1. INTRODUCTION

How to establish an effective supplier risk evaluation indicator system gets attention by domestic and foreign experts and scholars. Dickson (1996) [1] proposed 23 indicators to evaluate supplier risk. Wang (2010) [2] constructed an evaluation indicator system consisted of 6 first level indicators and 22 two level indicators. Liu (2010) [3] put forward seven non-financial indicators which represent suppliers' interests and realization method. Zhao (2011) [4] suggested to divide supplier risk indicators into supplier core competence evaluation indicators and performance evaluation indicators, and used ANP method to calculate the ranking and weight of supplier performance evaluation indicators. At present, most supplier risk evaluation is based on the establishment of risk indicators, which is obtained by traditional questionnaire and expert scoring method. However, to reduce so large amount of indicators, experts ought to constantly revise the scoring results and adjust to the judgment matrix to satisfy the consistency check. Although this excludes non conformance during indicator establishment, because of too many subjective factors, it can not avoid the serious one sidedness of individual decision maker. Based on the confirmed basic risk indicator system of supplier risk, make full use of vast historical data from SAP system, using rough set theory to reduce and optimize risk indicators, can access to relatively actual objective risk indicators, which can reflect the supplier risk more accurately.

2. ROUGH SET THEORY

Rough set theory is a data analysis theory proposed by Polish mathematician professor Z. Pawlak in 1980. It can effectively analyze and treat the imprecise, inconsistent and incomplete information, so as to find connotative knowledge, and reveal potential rules [5]. The main idea of rough set is according to the given knowledge of existing problem, classify and treat the actual measured data, and divide the domain of the problem into equivalence classes, to reduce the data and derive knowledge reduction and nuclear based on the remaining of key data information. It can evaluate the dependency between data and derive classification rules of concept. Rough set divides researched domain knowledge into knowledge expression system based on the indiscernible relationship, and uses utilization, lower approximation set to approximate to the described objects, through knowledge reduction, to get the most simplified knowledge [6, 7]. Rough set is based on knowledge and classification. However, knowledge is based on the ability to classify objects, and consists of classification mode. Different attribute knowledge description can get different classification. Classification is mainly used to produce category, these categories form knowledge module, which is the classification of class [8].

3. REDUCTION AND OPTIMIZATION STEPS OF SUP-PLIER RISK INDICATORS BASED ON ROUGHT SET

As a combination of qualitative analysis and quantitative tool, rough set can find rules from complicated and seemingly disordered data. At the same time, as a successful enterprise management system, SAP system provides massive basis data to find such rules. During the process of reduction and optimization of supplier risk indicators, based on the large amount of historical data in SAP system as a sample, our research use rough set theory to find rules. The specific steps are as follows:

Step 1: Set the range of each indicator according to the relationship between original data, and retrieve data from SAP database conforms to the condition, so as to establish knowledge set.

	Value of Risk Indicator																							
Supplier No	A1 %	A2 %	A3 %	A4 %	A5	A6	A7 %	B1 %	B2 %	B3	B4	B5 %	B6 %	B7 %	C1	C2 %	C3 %	C4	C5	D1 %	D2	D3	D4	Result
1	0.3	0.4	92	100	2	1	98	0.3	1.5	1	3	10	15	0.5	11	0.6	0	1	2	95	5	1	95	1
2	0.2	0.2	93	99	8	1	96	0.9	1.5	1	3	10	15	0.5	14	0.8	0	1	2	95	5	1	95	1
3	0.2	0.3	91	93	5	1	95	0.2	1.5	1	3	10	15	0.5	15	0.4	0	1	2	95	5	1	95	2
4	0.4	0.1	94	94	7	1	93	0.7	1.5	1	3	10	15	0.5	21	0.3	0	1	2	95	5	1	95	1
5	0.8	0.9	88	89	11	1	87	1.4	1.5	1	3	10	15	0.5	17	0.1	0.6	1	2	95	5	1	95	2
6	0.7	0.8	86	87	14	1	82	2.3	1.5	1	3	10	15	0.5	23	0	0.7	1	2	95	5	1	95	2
7	1.2	0.6	85	88	12	1	89	3.8	1.5	1	3	10	15	0.5	28	0	0.6	1	2	95	5	1	95	2
8	1.3	0.7	89	89	13	1	83	3.2	1.5	1	3	10	15	0.5	29	0	0.8	1	2	95	5	1	95	3
9	0.9	1.2	87	86	16	1	86	3.7	1.5	1	3	10	15	0.5	22	0	0.8	1	2	95	5	1	95	2
10	0.6	1.3	88	89	19	1	81	2.9	1.5	1	3	10	15	0.5	26	0	0.9	1	2	95	5	1	95	3
11	1.4	1.2	85	85	18	1	86	1.9	2.5	1	3	10	15	0.5	16	0	0.7	1	2	95	5	1	95	2
12	1.3	0.8	86	89	17	1	82	2.7	2.5	1	3	10	15	0.5	18	0	0.6	1	2	95	5	1	95	2
13	0.8	0.9	87	88	22	1	89	2.9	2.5	1	3	10	15	0.5	19	0	0.9	1	2	95	5	1	95	2
14	0.6	0.6	88	85	25	1	81	2.1	2.5	1	3	10	15	0.5	28	0	1	1	2	95	5	1	95	2
15	1.1	0.7	89	86	28	1	85	2.3	2.5	1	3	10	15	0.5	12	0	1	1	2	95	5	1	95	2
16	1.3	0.8	86	89	38	1	80	2.6	2.5	1	3	10	15	0.5	17	0	1	1	2	95	5	1	95	3
17	0.9	1.4	89	87	21	1	88	2.8	2.5	1	3	10	15	0.5	13	0	1	1	2	95	5	1	95	2
18	0.6	1.3	86	89	36	1	83	2.7	2.5	1	3	10	15	0.5	20	0	1	1	2	95	5	1	95	3
19	1.4	1.5	87	88	33	1	87	2.4	2.5	1	3	10	15	0.5	24	0	1	1	2	95	5	1	95	2
20	1.5	1.1	89	89	47	1	82	3.5	2.5	1	3	10	15	0.5	25	0.5	1	1	2	95	5	1	95	3
21	2.6	3.2	86	84	61	2	79	6.3	6.5	2	2	0	10	5	6	1.2	1.6	2	3	90	3	1	95	2
22	2.8	3.3	89	83	59	2	76	7.1	5.5	2	2	0	10	5	8	1.5	5.5	2	3	90	3	1	95	3
23	3.2	2.7	82	84	69	2	78	5.7	5.5	2	2	0	10	5	3	1.1	1.5	2	3	90	3	1	95	3
24	4.3	2.9	83	82	72	2	79	5.9	7.5	2	2	0	10	5	2	1.3	6	2	3	90	3	1	95	4
25	6.2	7.8	79	79	66	2	81	10.6	15	2	1	0	15	10	30	4.5	6	2	3	90	1	1	95	4

Table 1. SAP historical data according to risk early warning indicators for suppliers of common centrifugal pump accessories.

Step 2: According to evaluation results of users' preferences, form the decision table based on evaluation classification.

Step 3: Get reduction rules based on rough set theory, according to the confidence and support of rules to convert the rules into fuzzy constraint with priority, and then generate decision rules.

Step 4: Calculate the importance degree and weight for each indicator.

4. CONSTRUCTION AND PRE-TREATMENT OF DECISION TABLE

SAP system provides rich sample data for the comprehensive construction of indicator system. Sample data can be used to build decision table of supplier risk warning indicators, then use rough set theory to complete the reduction and optimization of each indicator. The optimized indicator system reduces the demand of information, and after repeated verification of actual data, it can be more accurately to reflect the true state of suppliers. To do that, first of all, it adopted the 23 indicators in literature [4] which was established by ANP method, and still took the common centrifugal pump parts suppliers as.

Some of the supplier risk indicator values in Table 1 are qualitative and others are quantitative. In accordance with the information reduction requirements for decision table, all of the attribute values need to be discredited. The meaning of the indicator values and their discretion rules are shown in Table 2.

Indicator No	Indicator Name	Grade of Discretion	Description
A1	Product quality complaint rate	1-5	1-[0, 0.5%];2-(0.5%-1%];3-(1%-1.5%];4-(1.5%-5%];5-More than 5%
A2	Return rate due to quality	1-5	1-[0, 0.5%];2-(0.5%-1%];3-(1%-1.5%];4-(1.5%-5%];5-More than 5%
A3	Product certification pass rate	1-5	1-No less than 90%; 2-[85%-90%); 3-[80%-85%); 4-[70%-80%); 5-Less than 70%
A4	Rate of delivery in accordance with order number	1-5	1-No less than 95%; 2-[90%-95%); 3-[85%-90%); 4-[80%-85%); 5-Less than 80%
A5	Ranking of performance	1-5	1-Top 10; 2-(10-20]; 3-(20-50]; 4-(50, 80]; 5-After 80
A6	Product grade	1-4	1-Excellent; 2-Good; 3-Moderate: 4-Other
A7	Reached rate of product standard life	1-5	1-No less than 90%;2-[85%-90%);3-[80%-85%);4-Less than 80%
B1	Price parity	1-5	Higher than average:1-No more than 0.5%;2-(0.5%-1%]; 3-(1%-5%]; 4- (5%-10%]; 5-More than 10%
B2	Price discount	1-3	1-No less than 10%; 2-[5%-10%); 3-[1%-5%); 4-Less than 1%
В3	Freight fee	1-3	1-Seller; 2-Arrive station; 3-Buyer
B4	Terms of payment	1-3	1-Installment; 2-Delivery pay; 3-Prepay
В5	Prepayment ratio of the order	1-5	1-[0, 5%];2-(5%-10%];3-(10%-15%];4-(15%-20%];5-More than 20%
B6	Payment ratio for quality warranty	1-5	1-[0, 5%];2-(5%-10%];3-(10%-15%];4-(15%-20%];5-More than 20%
B7	Financial fee	1-5	1-[0, 1%];2-(1%-3%];3-(3%-5%];4-(5%-10%];5-More than 10%
C1	Ranking of arrival time	1-5	1-Top 10;2-(10-30];3-(30-50];4-(50, 80];5-After 80
C2	Ratio of delayed delivery	1-5	1-[0, 1%];2-(1%-3%];3-(3%-5%];4-(5%-10%];5-More than 10%
C3	Breach of contract	1-5	1-[0, 0.5%];2-(0.5%-1%];3-(1%-5%];4-(5%-10%];5-More than 10%
C4	Delivery method	1-4	1-Motor transport; 2-Rail transport; 3-Pipeline transport; 4-Other
C5	Producing area and the distance	1-4	1-City; 2-Province; 3-Domestic; 4-Oversea
D1	Service prompt rate after sale	1-5	1-No less than 95%; 2-[90%-95%); 3-[85%-90%); 4-[80%-85%); 5-Less than 80%
D2	Guarantee period	1-4	1-No less than 5 years;2-[3-5);3-[1-3);4-In one year
D3	Communication skills of busi- ness man	1-4	1-Excellent; 2-Good; 3-Moderate; 4-Other
D4	Regular visit	1-5	1-No less than 95%; 2-[90%-95%); 3-[85%-90%); 4-[80%-85%); 5-Less than 80%

Table 2.	Meaning and	l discretion ru	les of indicator	values.
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According to the rules of discretion in Table 2, adding evaluation result of excellent, good, moderate, poor (R1, R2, R3, R4), the statistical data of decision table in Table 1 is simplified as shown in Table 3.

5. REDUCTION AND OPTIMIZATION OF SUPPLIER RISK INDICATORS

5.1. Divide Equivalence Classes for Each Risk Indicator

According to decision Table **3**, the equivalence classes are divided as follows:

 $\begin{array}{l} U/C= \{\{u1\}, \{u2\}, \{u3\}, \{u4\}, \{u5\}, \{u6\}, \{u7\}, \{u8\}, \\ \{u9\}, \{u10\}, \{u11\}, \{u12\}, \{u13\}, \{u14\}, \{u15\}, \{u16\}, \\ \{u17\}, \{u18\}, \{u19\}, \{u20\}, \{u21\}, \{u22\}, \{u23\}, \{u24\}, \\ \{u25\}\} \end{array}$

U/D= {{u1, u2, u4}, {u3, u5, u6, u7, u9, u11, u12, u13, u14, u15, u17, u19, u21}, {u8, u10, u16, u18, u22, u23, u20}, {u24, u25}}

 $\begin{array}{l} U/\{C-\{A1\}\} = \{\{u1\}, \{u2\}, \{u3\}, \{u4\}, \{u5, u7\}, \{u6, u8\}, \{u9, u11\}, \{u10\}, \{u12\}, \{u13, u15\}, \{u14, u16\}, \{u17, u19\}, \{u18, u20\}, \{u21\}, \{u22\}, \{u23\}, \{u24\}, \{u25\}\} \end{array}$

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Table 3.Simplified decision table.

Domain										Co	nditio	n Att	ribute	es C										Decision Attributed
U	A1	A2	A3	A4	A5	A6	A7	B1	B2	B3	B4	B5	B6	B 7	C1	C2	C3	C4	C5	D1	D2	D3	D4	
1	1	1	1	1	1	1	1	1	3	1	3	2	3	1	2	1	1	1	2	1	1	1	1	R1
2	1	1	1	1	1	1	1	2	3	1	3	2	3	1	2	1	1	1	2	1	1	1	1	R1
3	1	1	1	2	1	1	1	1	3	1	3	2	3	1	2	1	1	1	2	1	1	1	1	R2
4	1	1	1	2	1	1	1	2	3	1	3	2	3	1	2	1	1	1	2	1	1	1	1	R1
5	2	2	2	3	2	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
6	2	2	2	3	2	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
7	3	2	2	3	2	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
8	3	2	2	3	2	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R3
9	2	3	2	3	2	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
10	2	3	2	3	2	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R3
11	3	3	2	3	2	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
12	3	3	2	3	2	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
13	2	2	2	3	3	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
14	2	2	2	3	3	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
15	3	2	2	3	3	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
16	3	2	2	3	3	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R3
17	2	3	2	3	3	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
18	2	3	2	3	3	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R3
19	3	3	2	3	3	1	2	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R2
20	3	3	2	3	3	1	3	3	3	1	3	2	3	1	2	1	2	1	2	1	1	1	1	R3
21	4	4	2	4	4	2	4	4	2	2	2	1	2	3	1	2	3	2	3	2	2	1	1	R2
22	4	4	2	4	4	2	4	4	2	2	2	1	2	3	1	2	4	2	3	2	2	1	1	R3
23	4	4	3	4	4	2	4	4	2	2	2	1	2	3	1	2	3	2	3	2	2	1	1	R3
24	4	4	3	4	4	2	4	4	2	2	2	1	2	3	1	2	4	2	3	2	2	1	1	R4
25	5	5	4	5	4	2	3	5	1	2	1	1	3	4	2	3	4	2	3	2	3	1	1	R4

 $\begin{array}{l} U/\{C-\{A2\}\}= \{\{u1\}, \ \{u2\}, \ \{u3\}, \ \{u4\}, \ \{u5, \ u9\}, \ \{u6, \ u10\}, \ \{u7, \ u11\}, \ \{u8, \ u12\}, \ \{u13, \ u17\}, \ \{u14, \ u18\}, \ \{u15, \ u19\}, \ \{u16, \ u20\}, \ \{u21\}, \ \{u22\}, \ \{u23\}, \ \{u24\}, \ \{u25\}\} \end{array}$

 $\begin{array}{l} U/\{C-\{A3\}\}= \{\{u1\}, \{u2\}, \{u3\}, \{u4\}, \{u5\}, \{u6\}, \\ \{u7\}, \{u8\}, \{u9\}, \{u10\}, \{u11\}, \{u12\}, \{u13\}, \{u14\}, \\ \{u15\}, \{u16\}, \{u17\}, \{u18\}, \{u19\}, \{u20\}, \{u21, u23\}, \\ \{u22, u24\}, \{u25\}\} \end{array}$

 $\begin{array}{l} U/\{C-\{A4\}\} = \{\{u1, u3\}, \{u2, u4\}, \{u5\}, \{u6\}, \{u7\}, \\ \{u8\}, \{u9\}, \{u10\}, \{u11\}, \{u12\}, \{u13\}, \{u14\}, \{u15\}, \\ \{u16\}, \{u17\}, \{u18\}, \{u19\}, \{u20\}, \{u21\}, \{u22\}, \{u23\}, \\ \{u24\}, \{u25\}\} \end{array}$

 $\begin{array}{l} U/\{C-\{A5\}\} = \{\{u1\}, \{u2\}, \{u3\}, \{u4\}, \{u5, u13\}, \{u6, u14\}, \{u7, u15\}, \{u8, u16\}, \{u9, u17\}, \{u10, u18\}, \{u11, u19\}, \{u12, u20\}, \{u21\}, \{u22\}, \{u23\}, \{u24\}, \{u25\} \end{array}$

 $U/{C-{A6}} = U/{C-{A5}}$

 $\begin{array}{l} U/\{C-\{A7\}\}=\{\{u1\}, \ \{u2\}, \ \{u3\}, \ \{u4\}, \ \{u5, \ u6\}, \ \{u7, \ u8\}, \ \{u9, \ u10\}, \ \{u11, \ u12\}, \ \{u13, \ u14\}, \ \{u15, \ u16\}, \ \{u17, \ u18\}, \ \{u19, \ u20\}, \ \{u21\}, \ \{u22\}, \ \{u23\}, \ \{u24\}, \ \{u25\}\} \end{array}$

 $\begin{array}{l} U/\{C-\{B1\}\}= \{\{u1,\ u2\},\ \{u3,\ u4\},\ \{u5\},\ \{u6\},\ \{u7\},\ \{u8\},\ \{u9\},\ \{u10\},\ \{u11\},\ \{u12\},\ \{u13\},\ \{u14\},\ \{u15\},\ \{u16\},\ \{u17\},\ \{u18\},\ \{u19\},\ \{u20\},\ \{u21\},\ \{u22\},\ \{u23\},\ \{u24\},\ \{u25\}\} \end{array}$

 $U/{C-{B2}} = U/{C-{A4}}$

- U/{C-{B3}}=U/C U/{C-{B4}}=U/C U/{C-{B5}}=U/C
- $U/{C-{B6}} = U/C$

 $U/{C-{B7}} = U/C$

$$U/{C-{C1}} = U/C$$

 $U/{C-{C2}} = U/C$

 $\begin{array}{l} U/\{C-\{C3\}\}=\{\{u1\}, \{u2\}, \{u3\}, \{u4\}, \{u5\}, \{u6\}, \{u7\}, \\ \{u8\}, \{u9\}, \{u10\}, \{u11\}, \{u12\}, \{u13\}, \{u14\}, \{u15\}, \\ \{u16\}, \{u17\}, \{u18\}, \{u19\}, \{u20\}, \{u21, u22\}, \{u23, u24\}, \\ \{u25\}\} \end{array}$

 $U/{C-{C4}} = U/C$ $U/{C-{C5}} = U/C$ $U/{C-{D1}} = U/C$ $U/{C-{D2}} = U/C$

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 $U/{C-{D3}} = U/C$

$$U/{U-{D4}} = U/C$$

5.2. Calculate Positive Domain and Obtain the Importance Degree of Each Indicator

According to rough set theory and the divided equivalence classes above, calculate positive domain and the importance degree of each indicator as follows:

posC(D)= {u1, u2, u3, u4, u5, u6, u7, u8, u9, u10, u11, u12, u13, u14, u15, u16, u17, u18, u19, u20, u21, u22, u23, u24, u25}

 $posC-{A1}(D)= \{ u1, u2, u3, u4, u5, u7, u9, u10, u11, u12, u13, u15, u17, u18, u19, u20, u21, u22, u23, u24, u25)\}\neq posC(D), A1 is an essential condition, its importance degree: <math>25/25-21/25=4/25$

 $posC-{A2}(D)= \{u1, u2, u3, u4, u5, u7, u9, u11, u13, u15, u16, u17, u19, u20, u21, u22, u23, u24, u25\}\neq posC(D), A2 is an essential condition, its importance degree: 25/25-19/25=6/25$

 $posC-{A3}(D)= \{u1, u2, u3, u4, u5, u6, u7, u8, u9, u10, u11, u12, u13, u14, u15, u16, u17, u18, u19, u20, u25\} \neq posC(D), A3 is an essential condition, its importance degree: <math>25/25-21/25=4/25$

posC-{A4}(D)= {u2, u4, u5, u6, u7, u8, u9, u10, u11, u12, u13, u14, u15, u16, u17, u18, u19, u20, u21, u22, u23, u24, u25} \neq posC(D), A4 is an essential condition, its importance degree: 25/25-23/25=2/25

 $posC-{A5}(D)= \{u1, u2, u3, u4, u5, u6, u7, u8, u9, u10, u11, u13, u14, u15, u16, u17, u18, u19, u21, u22, u23, u24, u25} \neq posC(D)$, A5 is an essential condition, its importance degree: 25/25-23/25=2/25

 $posC-{A6}(D)=posC-{A5}(D)$, A6 is an essential condition, its importance degree: 2/25

 $posC-{A7}(D)=$

{u1, u2, u3, u4, u5, u6, u11, u12, u13, u14, u21, u22, u23, u24, u25} \neq posC(D), A7 is an essential condition, its importance degree: 25/25-15/25=10/25

 $posC-{B1}(D)=$

{u1, u2, u5, u6, u7, u8, u9, u10, u11, u12, u13, u14, u15, u16, u17, u18, u19, u20, u21, u22, u23, u24, u25}

 \neq posC(D), B1 is an essential condition, its importance degree: 25/25-23/25=2/25

 $posC-{B2}(D)=posC-{A4}(D)$, B2 is an essential condition, its importance degree: 2/25

posC-{C3}(D)= {u1, u2, u3, u4, u5, u6, u7, u8, u9, u10, u11, u12, u13, u14, u15, u16, u17, u18, u19, u20, u23, u24, u25} \neq posC(D), C3 is an essential condition, its importance degree: 25/25-23/25=2/25

The other indicators are unessential conditions and can be reduced.

5.3. Normalization and Weight of Indicators

Within Normalized each essential indicator, the weight of reduced and optimized risk warning indicators are shown in Table **4**.

 Table 4.
 Weight table of reduced and optimized early warning indicators by rough set.

Indicator No	Indicator Name	Weight
A1	Product quality complaint rate	0.058824
A2	Return rate due to quality	0.176471
A3	Product certification pass rate	0.117647
A4	Rate of delivery in accordance with order number	0.058824
A5	Ranking of performance	0.058824
A6	Product grade	0.058824
A7	Reached rate of product standard life	0.294118
B1	Price parity	0.058824
B2	Price discount	0.058824
C3	Breach of contract	0.058824

SUMMARY AND CONCLUSION

Considering the influence of subjective factors during the construction of indicator system using traditional method, based on initial determined indicators derived from traditional experience, combined with extracted historical data in enterprises' SAP system, this paper used rough set theory to reduce and optimize the original risk indicator system and lower the system dimension. Then it recalculated their weights and optimized the risk evaluation rules. Because of the using of newly real data from suppliers' business, this method can establish risk early warning indicator system for each material catalog dynamically and in real time. Indicator system which is optimized and validated can reflect the real state of suppliers more accurately with less information requirement.

CONFLICT OF INTEREST

The authors confirm that this article content has no conflict of interest.

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