

SOM Network Based Clustering Analysis of Real Estate Enterprises

Jinhong Zhu, Sheng Liu*

School of Management, Shanghai University of Engineering Science, Shanghai, China Email: purplelight713@sina.com, *ls6601@163.com

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Abstract

For Real Estate industry which has many types of enterprises, how to carry on the effective clustering analysis has become a problem that needs to solve. This paper first theoretically elaborates the SOM network, and then pretreats the data with SOM network, which has the ability to deal with the high dimensional and complex data. Finally it uses the clustering function of SOM neural network to make clustering analysis and comparison of Real Estate companies which are listed in Shanghai and Shenzhen stock market. The clustering analysis results based on SOM are displayed in two-dimensional graphics, showing intuitively and comprehensively of the financial situation of each enterprise.

Keywords

Real Estate Enterprises; Financial Situation; SOM Neural Network; Clustering Analysis

1. Introduction

In recent years, the Real Estate Enterprises in China is affected by domestic and international external environmental factors as an important mainstay of the market economy, such as changes in market, the economic environment, social and cultural environment and policy changes. All these ever-changing factors are difficult to accurately predict that they inevitably bring about a direct or indirect impact on the financial management of Real Estate Enterprises [1]. These factors have a profound impact on companies when they deal with various financial relationships and activities. There are uncertain differences between financial incomes and expected incomes which have had a significant impact on the survival, profitability and sustainable development of the enterprises. Nowadays, China's Real Estate Enterprises are in a very competitive market environment of both domestic and international arena with the growing competition in domestic and international market and the in-

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^{*}Corresponding author.

creasing complexity of the market competition environment [2].

Development of enterprises has a certain periodicity and the Real State Enterprises are no exception. They also experience the development cycle of production, development, prosperity and recession. The cycle of national economic development and the level of it have a great influence on Real Estate Enterprises. Business investment is still lacking in normative behaviors in China's Real Estate Enterprises because of the late start of China's Real Estate industry [3]. They are facing with a series of financial risks with suffering from the changes of global economic environment and the impact of domestic macro-control policies and other factors. Financial risk management and control apparently have become significant problems in the process of Real Estate Enterprises' development [4].

Real Estate industry is a typical capital-intensive industry which needs to invest a lot of money. With the continuous improving cost of land and construction, the long period of Real Estate projects' exploit, the long time capital takes up and the poor ability to adjust projects, you need to invest a lot of money to operate. Some Real Estate developers are lacking in awareness of the financial budget analysis. They don't have a full analysis of the budget or in-depth investigation and study on the cost of land, financial operations, economic rate of return and other issues which the Real Estate development projects are facing. They have declinational judgment on the priority of use of funds, and are lack of overall planning for the use of the entire project funds, thus causing blind operation on funds management. Some developers are lack of planning for borrowing. They apply for excessive loan to banks which is beyond the ability of themselves or borrow big money from other companies. All these increase the financial risks of Real Estate Enterprises [5].

As the global economy continues to grow and prosper, the enterprise financial evaluation system needs faster and more accurate processing ability on huge data. The improvement and integration of artificial intelligence method is the mainstream trend of the current financial evaluation methods. Numerous studies show that the neural network has relatively strong predictive ability and a high prediction accuracy rate in financial early warning. The clustering function of SOM neural network makes a variety of models cluster automatically in accordance with certain distances. Thus, it does not require the training sets that other clustering methods need and avoids the subjective interference of classification. The clustering analysis forms richer classes of species and provides a feasible technical method of analysis and evaluation of financial situation which is comprehensive and structured [6].

2. SOM Neural Network

According to the study in biology, in many living organisms' pallium, a part of input signals of neurons are from the sensory organization or the external input signals of other areas, while the other parts are from the feedback signals of the same region. The weighted sum of the input signals received by each neuron has a function of feature detection, and the lateral feedback connections has a function of excitation or inhibition, which depends on the different distance with each neuron [7]. Teuvo Kohonen put forward self-organization mapping network model in 1981 according to this characteristic of biological neurons. He thought that neurons which are in different regions of space have different division of labor, and a neural net will be divided into different response areas while each area has a different response characteristic when to accept outside input mode. Self-organization mapping network is a kind of classification method of unsupervised learning [8]. It can avoid some disadvantages which the traditional clustering methods are difficult to overcome, and can use its strong learning function and good self-organization, adaptive and robustness to provide information processing methods for the study on nonlinear classification, which has been successfully applied to the spatial pattern recognition problems, text clustering, natural language processing and many other research fields [9].

One of the typical characteristics of SOM network is its ability to make the n-dimensional space data input mapped to a lower dimension (usually one or two dimensional) output. It finds out and extracts its inherent characteristics from large amount of input data through the self-organization of network structure. And it shapes a topological graph of the input data distribution in the vector space of network output node weight, which reflects a certain distribution of input data, while keeping the original topological relationship of data which can be used for cluster analysis. So SOM is a cluster analysis technology which has a visual result. It can reveal the interrelationship between the indexes in high dimensional data by graphics. SOM neural network has a function of cluster analysis, making all kinds of modes cluster automatically according to a certain distance rule. So that it not only does not require the training sets which other classification methods need, but also avoids the subjective interference of classification, and the cluster analysis forms more abundant category types [10]. Its difference

with other types of neural networks is: it reflects the classification results not by the state vector of a neuron or a network, but by a number of neurons simultaneously (in parallel). This kind of feature mapping neural network makes it possible for the space distribution of connection weights vector to reflect the input mode's statistical characteristics by iterative learning of the input mode [11].

SOM network is composed of input layer and output layer. Each neuron in the input layer connects with every neuron in the output layer by weights. The input layer neurons are arranged in a one-dimensional form, and the number of input neurons is determined by the number of components in the input vectors [12]. The output layer neurons are generally arranged in a one-dimensional or two-dimensional form, as shown in **Figure 1**.

The training process of SOM algorithm is divided into 5 steps:

1) Establish networks according to the dimensions and the number of training samples, and give small weights to connection weights of M input neurons to the output neurons and complete the initialized weight matrix W [13];

2) Set the input space is of dimension m: $X = [X_1, X_2, \dots X_m];$

3) Calculate the Euclidean distance between the input samples and each output neuron j:

$$d_{j} = \left\| X - W_{j} \right\| = \sqrt{\sum_{i=1}^{n} \left[X_{i}(t) - W_{ij}(t) \right]^{2}}$$
$$d_{j} = \left\| X - W_{j} \right\| = \left[X_{i}(t) - W_{ij}(t) \right]$$

Select the best matching unit which has the minimum distance of neurons. And the weight of the node *j* is:

$$W_{j} = \left[W_{j1}, W_{j2}, W_{j3}, \cdots, W_{jm}\right];$$

4) Determine the area function, and correct the weight according to the type below:

$$W_{ij}(t+1) = W_{ij}(t) + \eta(t) \left[X_i(t) - W_{ij}(t) \right]$$

Among them, η is a gain which declines to zero gradually with time, and generally $\eta(t) = 0.2 \cdot (1-t)$ or $\eta(t) = 0.2 \cdot (1-t/1000)$

5) Normalized output:

$$O_k = f\left(\min\left\|X - W_j\right\|\right)$$

In the training process, the neurons participate in each other's competition, and the neuron node which has a maximum output is the winner. The winning node has the capacity to restrain other competitors and activate its neighboring nodes. But because only the winning node is the best match to the input graphs, only the winner is allowed to output, and only the weights of the winner and its adjacent nodes are allowed to be adjusted [14]. That is to say SOM network imitates the distribution of the input graphs, or to say it can extract the characteristics of the input graphs, and classify the input graphs according to the similar characteristics, representing by a



winning node [15].

From the above discussion that, the space distribution of neural weight vectors can reflect the statistical characteristics of input vectors for the trained self-organization mapping network.

This paper explores the self-organizing mapping network model which is in the application of the Real Estate enterprise financial analysis, and classifies the Real Estate companies' financial condition according to clustering analysis by SOM model and the actual data. The validity of the model is verified by experimental results which are satisfactory.

3. Clustering Indexes Selection

This paper uses financial indexes of Real Estate companies which are listing in Shanghai and Shenzhen stock market to research financial situation of them. The rational and scientific financial risk indexes play an important role in description of financial situation of a company. Quantitative indexes include the cash flow capacity, solvency, profitability, operation ability, development capacity [16]. Among them, the cash flow capacity is mainly used to reflect the cash flows that brought by business operating activities of enterprises [17]; Solvency is the ability of companies to repay all kinds of maturing debt that can be divided into short-term solvency and long-term solvency [18]; Profitability is company's ability to obtain profits, reflecting the company's operating performance and financial structure [19]; Operation ability refers to an enterprise's capability of configuring and using its limited resources; Development capacity refers to the enterprise's prospect, trend and potential of its production and business operation activities in the future [20].

This paper selects 6 financial risk indexes in order to make a comprehensive analysis on the financial condition of Real Estate Enterprises, as shown in **Table 1**.

4. Experiment and Result Analysis

This paper selects the Shanghai and Shenzhen listed Real Estate companies of the 27 financial indexes, and data as of June 30, 2013.

The main parameter settings of SOM clustering algorithm are as follows:

1) Input layer: The input layer parameters are the 6 indicators and 38 stock codes of Real Estate companies which are listed, summing to data of 6×38 . As neural network training samples, the value is entered after dealing with the normalized can increase the area of the change, which greatly improves the training of convergence [21]. Conversion of essential data is as following:

$$b_i = \frac{\alpha_i - \alpha_{\min}}{\alpha_{\max} - \alpha_{\min}}$$

2) Output Layer: The output layer is a two-dimensional plane space, constituted by four neurons (2 * 2).

3) Use Matlab to deal with data, and the clustering result is shown in Figure 2.

As shown in **Figure 2**, the enterprises whose ticker symbols are in the first and second layer develop more balanced, especially the ability to grow. And their profitability is the strongest. They are worth the investment because of their high safety degree of debt loans and preferable financial flexibility. The enterprises whose ticker symbols are in the third and fourth layer can be a small investment because they are mainly distributed to high security degree of creditor's rights and their operational efficiency and solvency are in general. The enterprises whose stock codes are at Layer 5 and Layer 6 can be an appropriate investment because of their general profitability and growth ability. And they have higher solvency and a higher level of debt loans security. The enterprises whose stock codes are at Layer 7 and Layer 8 are overall the worst. They can be appropriate investments; ticker symbol in the fifth floor of the company's overall worst. They have the poorest profitability, growth ability and financial flexibility although their debt paying ability is in general. So they are recommended not to be invested.

5. Conclusion

SOM network model is a new method of clustering analysis, which can be used to analyze the financial situation of Real Estate companies. It has not only the characteristics of fast convergence, small amount of calculation and low computational complexity, but also the characteristics of its self-learning, self-adaptability and strong

Number	Stock Code	Liquidity Ratio	Net Profit Margin on Sales	Total Assets Turnover	Earnings Per Share	Asset-liability Ratio	Return on Asse In Cash
a1	000002	1.914878	0.131544	0.380626	0.48	0.670017	0.067244
a2	000024	1.435508	0.119855	0.407494	0.3977	0.709909	0.246129
a3	000046	1.798057	0.173063	0.2376	0.96	0.617889	0.147289
a4	000534	1.326004	0.053367	0.29987	0.2828	0.677765	0.083083
a5	000014	4.283123	0.164104	0.111788	0.18	0.607049	-0.067527
a6	000043	6.822626	0.191101	0.219456	0.1	0.560379	-0.201905
a7	000511	1.888322	0.086667	0.401498	0.0565	0.512337	0.19942
a8	000537	1.209801	0.126035	0.361904	0.16	0.726146	0.155653
a9	000540	2.580682	0.280623	0.357243	0.8066	0.801839	-0.077738
a10	000608	1.300603	0.304647	0.146957	0.5	0.701256	0.168963
a11	000628	0.919686	0.012543	0.494542	0.053	0.943144	0.070949
a12	000638	1.494224	0.124837	0.654482	0.24	0.671678	0.147005
a13	000671	1.598108	0.081013	0.743797	1.08	0.623011	0.23622
a14	000711	2.18295	-0.033495	0.073644	0.01	0.544582	0.02219
a15	000736	3.17719	1.30844	0.040651	0.35	0.386379	-0.055939
a16	000897	1.874149	0.009907	0.233948	0.0102	0.681427	0.11886
a17	000965	1.727456	0.247559	0.175969	0.34	0.514705	0.16269
a18	002133	2.043243	0.125882	0.219479	0.33	0.726385	0.178627
a19	002244	1.718157	0.221683	0.17202	0.48	0.779288	-0.005211
a20	600048	2.314826	0.174351	0.320454	1.06	0.699863	-0.012744
a21	600052	1.502249	0.09985	0.50135	0.5	0.772454	0.075723
a22	600053	2.348969	0.029785	0.171316	0.04	0.711111	0.120521
a23	600082	4.773068	0.11216	0.26558	0.14	0.391763	-0.03607
a24	600215	3.407249	0.005593	0.158288	0.008	0.255525	0.081291
a25	600225	1.277079	0.094422	0.261592	0.32	0.87948	0.052691
a26	600239	3.782101	0.150406	0.385051	0.3578	0.618071	-0.332066
a27	600240	2.219439	0.095983	0.236736	0.1814	0.69003	-0.008942
a28	600266	1.855596	0.242521	0.266624	1.1471	0.674722	0.113208
a29	600322	2.336302	0.141891	0.118509	0.14	0.596637	0.05642
a30	600383	2.211707	0.15981	0.267011	0.78	0.696484	-0.051567
a31	600638	2.687847	0.319037	0.100065	0.285	0.542947	-0.003685
a32	600648	0.894669	0.05742	0.361708	0.35	0.777776	0.05544
a33	600663	1.3785	0.566867	0.187306	0.6062	0.423997	-0.055975
a34	600665	1.470336	0.104637	0.339791	0.193	0.735293	0.209844
a35	600675	2.243801	0.189868	0.345475	0.585	0.695883	0.00271
a36	600684	2.976838	0.117458	0.331622	0.33	0.562685	0.087508
a37	600890	1.960102	0.363531	0.252156	0.07	0.370151	0.214552
a38	900950	1.494947	0.094692	0.454023	0.4607	0.818658	0.096109

 Table 1. Financial analysis index system of real estate enterprises.



Figure 2. SOM data training result.

fault tolerance. All these make classification of intelligent available, and lead to satisfying results. Seen from the interior of Real Estate companies that you can realize the changes of financial situation of these companies in a timely manner, and compared with other scale-like enterprises, you can avoid the stagnation of various projects caused by the capital chain rupture in order to avoid the crisis of bankruptcy. Seen from an external perspective at all levels of the companies that the majority of shareholders can buy stocks according to the hierarchy of Real Estate Enterprises' financial condition to avoid the potential risk of loss caused, Banks can use this model for credit control of Real Estate Enterprises, which not only can reduce the risk of loans, but also laid the foundation for the stable operation of the national economy. In addition, external auditors can narrow the scope of audit objects and develop targeted audit procedures, thereby improving their ability to assess and reducing audit risks.

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