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# A Hybrid Portfolio Selection Model based on Grey Relational Analysis and Feature Selection: Evidence on Listed Firms in Tehran Stock Exchange

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ARTICLE INFO	ABSTRACT
Article history:	Background: A portfolio is a collection of investments held by an institution or private
Received 20 March 2014	individual. Portfolio selection is about choosing which assets and in what proportion
Received in revised form 20	will best respect the investor's preferences for achieving an expected return assuming a
April 2014	minimum risk Objective: write the main objective for your paper. Results: The
Accepted 15 May 2014	purpose of stock portfolio selection is how to allocate the capital to a large number of
Available online 1 June 2014	stocks in order to bring a most profitable return for investors. The main aim of this
	research is selection optimized portfolio using grey relational analysis (GRA) and
Keywords:	feature selection methods. First, this paper used of feature selection method to identify
grey relational analysis (GRA)	key indicators as criteria for recognition of proper firms. Second, uses of grey relational
feature selection portfolio	analysis (GRA) method to ranking of examined firms for selecting a basket of stocks
Tehran Stock Exchange	that has the best performance. 50 companies were selected as the sample among listed
	companies in Tehran Stock Exchange during 2000 to 2013. Conclusion: The Results of
	this study indicate that a grey relational analysis (GRA) technique is employed to
	specify an appropriate weighting of the selected stocks such that the portfolio rate of
	return is maximized. Finally, was found that, both of (GRA) and feature selection
	methods are fit.

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#### INTRODUCTION

Identifying the "best" portfolio of assets for an individual investor is one of the principal challenges in the world of finance. A portfolio is a collection of investments held by an institution or private individual. Portfolio selection is about choosing which assets and in what proportion will best respect the investor's preferences for achieving an expected return assuming a minimum risk. In the classical portfolio selection formulation provided by Markowitz (1952), the return is quantified by means of its expected value and the variance of the portfolio return is regarded as the risk of the investment. Markowitz' model (1952) uses the mean and variance of historical returns to measure the expected return and risk of a portfolio. Conventionally, such portfolio selection problems are solved with quadratic or linear programming models under the assumption that the asset weights in the portfolio are real numbers, which are difficult to implement. Specifically, each asset has its minimum transaction lot, while the solutions involve only real-number asset weights rather than asset trading units. For example, stocks might be traded at the unit one share, and mutual funds have their individual minimum trading amounts. Thus, the solution obtained by Markowitz' model must be integers to be applicable in practice. Other than Markowitz' model, (Speranza, 1996; Mansini, Speranza, 1997, 1999; Kellerer, et al, 2000) proposed their respective portfolio selection models. based on Konno and Yamazaki's mean absolute deviation (MAD) model (Konno, Yamazaki, 1991). Speranza, (1996) proposed a mixed integer program considering realistic characteristics in portfolio selection, such as minimum transaction lots and the maximum number of securities, and suggested a simple two-phase heuristic algorithm to solve the proposed integer program. Mansini and Speranza (1997,1999) showed that the portfolio selection problem with minimum transaction lots is an NPcomplete problem and proposed three heuristic algorithms to solve the problem. Based on the MAD model, Konno and Wijayanayake (2001) proposed an exact algorithm for portfolio optimization problems under concave transaction costs and minimum transaction lots. However, minimum transaction lots were not the major concern in their study. Later, Mansini and Speranza (2005) derived a mean safety model with side constraints from the MAD model, and proposed an exact algorithm to solve for portfolios under the consideration of transaction costs and minimum transaction lots. However, Markowitz' model is still the most widespread

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portfolio selection model. Solving the portfolio selection problem based on Markowitz' model and, simultaneously, considering minimum transaction lots are of practical significance. However, it appears that no methods in the past solving the portfolio selection problem with minimum transaction lots were based on Markowitz' model. Deng (1982) founded context of gray relations based on the theory of systems. This method consults togetherness among components of one system and reference series (Deng, 1988; Huang et al., 2008). This theory is used to solve ambiguous problems and the problems having disconnected and incomplete data. It provides satisfactory and popular outputs by a little data and with many changes in criteria. Gray theory, like fuzzy theory is an effective mathematic model to solve indefinite and ambiguous problems. This theory is used in many fields and has been utilized in the field of solving multi criterion decision making problems named gray relational analysis. Gray relational analysis being one of gray relation is used for solving complex relations between factors and variations for solving problems. Theory of Gray systems is an algorithm that analyzes nonlogical relations of one system members with a reference member and it includes capability of solving multi criterion decision making problems. In many research problems, such as pattern recognition, it is important to choose a group of set of attributions with more prediction information. That is, if the number of irrelevant or redundant features is reduced drastically, the running time of a learning algorithm is also reduced. Moreover, a more general concept can be yielded. Performing feature selection can lead to many potential benefits, which are facilitating data visualization and data understanding, reducing the measurement and storage requirements, reducing training and utilization times, defying the curse of dimensionality to improve prediction performances, etc. (Guyon, Elisseeff, 2005; Kim, 2006; Mladenic´, Grobelnik, 2003).

By using of grey relational analysis (GRA) and feature selection methods in this research having been a little focused and ignored in previous studies; we are going to introduce a hybrid model for optimal selection of portfolio. The purpose of stock portfolio selection is how to allocate the capital to a large number of stocks in order to bring a most profitable return for investors. Therefore, the main aim of this research is selection optimized portfolio using grey relational analysis (GRA) and feature selection methods. First, this paper used of feature selection method to identify key indicators as criteria for recognition of proper firms. Second, uses of grey relational analysis (GRA) method to ranking of examined firms for selecting a basket of stocks that has the best performance.

#### Literature:

The main researches that conducted in order to stock selection issue, used of some models such as: Markowitz, DEA, AHP, MCDM, Electra and TOPSIS. Furthermore, there are a few researches that used of gray relational analysis (GRA) for portfolio selection. This section reviews the literature of previous work that used of above models.

Beshkooh and Afshari (2012), used of AHP and GRA theory in Tehran stock exchange to portfolio selection and encountered some indicators such as: Stock Prices, Operating income ratio, Company size, Current Ratio, Inventory turnover, weighted average index. Tasi and Hsio (2012) investigated Genetic algorithm in Taiwan stock exchange to portfolio selection and used of some variables i.e. financial variables, Macroeconomic variables, Stock price information, Technical indicators.

Lin and liu (2006), surveyed Taiwanese investment funds Data on the basis of Markowitz, Fuzzy multiobjective decision making models and encountered some indicators such as: Risk, Monthly returns, Return covariance between assets per year. Wang (2003) investigated Fuzzy rough set model in Taiwan stock exchange to portfolio selection and used of some variables such as: Stock Prices, Trends Stock, Mining Rules, Details Buy / Sell. Huang (2012) used of Genetic algorithms, Support vector regression in Taiwan stock exchange and applied of some indicator i.e. Profitability Growth, Share prices, Liquidity. Bulgurca (2012) investigated TOPSIS model in Istanbul stock exchange to portfolio selection and used of some variables i.e. Current Ratio, Total debt ratio, Debt, Current assets, Turnover,

Turnover of fixed assets, Profit (loss), net. Yi and *et al.* (2010) surveys 8 firms of chines steel industry and implies of AHP and GRA models. Furthermore, they used of several indicators such as: Profitability, Profitability to shareholders, Asset Management. Chen and hirasawa (2010) used of GRA and Genetic

Network Programming in Tokyo stock exchange to portfolio selection and encountered some indicators such as: Return, Budget and Profitability. See summary of the literature in table 1.

On the basis above literature review, this paper used of a hybrid model to portfolio selection with using of GRA and feature selection methods. Therefore, current study has a new point of view to portfolio selection problem. In this paper used some indicators such as: company size, ROE, ROA, P/BV, stock price, Current Ratio, Liquidity ratio, Operating Profit, Net profit to sales.

#### Methods:

#### Grey relational analysis (GRA):

Grey system theory, which was presented by Deng in 1982, is a new mathematical theory that was born by idea of grey set theory. It is one of efficient methods that are used to solve uncertainty and vogue problems

under discrete data and incomplete information. In grey system theory, according to information degree, if the system information is fully known, the system is called a white system, while the system information is unknown; it is called a black system. We give concept a grey system as Figure 1.

9	Author (s)	Indicator (s)	Method (s)	Nature of decision
1	Beshkooh and Afshari	Stock Prices	AHP	Non Fuzzy
	(2012)	Operating income ratio	GRA	
		Company size		
		Current Ratio		
		Inventory turnover		
		weighted average index		
2	Tasi and Hsio	Financial variables		Non Fuzzy
	(2012)	Macroeconomic variables	Genetic algorithm	-
		Stock price information	-	
		Technical indicators		
3	Lin and liu	Risk	Markowitz	Fuzzy
	(2006)	Monthly returns	Fuzzy multi-objective	
		Return covariance between assets	decision making	
		per year	Ũ	
4		Stock Prices		Fuzzy
	Wang	Trends Stock	Fuzzy rough set	•
	(2003)	Mining Rules		
		Details Buy / Sell		
5	Huang	Profitability	Genetic algorithms	Non Fuzzy
	(2012)	Growth	Support vector	
		Share prices	regression	
		Liquidity	_	
6		Current Ratio		Non Fuzzy
	Bulgurca	Total debt ratio	TOPSIS	
	(2012)	Debt	101515	
		Current assets		
		Turnover		
		Turnover of fixed assets		
		Profit (loss), net		
7	Yi and <i>et al</i> .	Profitability	AHP	Non Fuzzy
	(2010)	Profitability to shareholders	GRA	
		Asset Management		
8	Chen and hirasawa	Return	GRA	Non Fuzzy
	(2010)	Budget	Genetic	
	· · ·	Profitability	Network Programming	

# Grey system



Fig. 1: the concept of a grey system.

A system with partial information known and partial information unknown is grey system. The theory includes five major parts, which include grey prediction, GRA, grey decision, grey programming, and grey control (Tzeng and Huang, 2011).

The GRA is an important approach of grey system theory in the application of estimating alternatives through decision attributes. In GRA, the data that contain same features are regarded as a sequence. As a tool of quantitative and qualitative analyses, the GRA can be used to measure the relation between two sequences by calculation their correlative degrees, which is called grey relational grade (GRG). The GRG is expressed by a scalar between 0 and 1. Up to now, the method used in resolving MADM issues (Li *et al.*, 2009). The main idea of GRA is first transforming the performance of all alternatives into a comparability sequence. This step is called grey relational generating. According to these sequences, a reference sequence (ideal target sequence) is defined. Then, grey relational coefficient between all comparability sequences and the reference sequence and every comparability sequences is calculated. If a comparability sequence translated from an alternative has the highest GRG between the reference sequence and that alternative will be the best choice (Kuo *et al.*, 2008). The procedures of GRA are shown in Figure 2. The detail of the proposed GRA procedure is presented below:

Step 1 Determination of comparability sequences and reference sequence. Let  $X1 \sim Xm$  be comparability sequences. Then comparability matrix *D* is formulated as follows:

$$D = \begin{bmatrix} x_{11} & \cdots & x_{m^1} \\ \vdots & \ddots & \vdots \\ x_{1^n} & \cdots & x_{m^n} \end{bmatrix}$$

Let  $y_1 = (y_1(1), y_1(2), \dots, y_1(n))^T$  be reference sequence. The GRGs between the comparability sequences and reference sequence denote the relation degrees between the Technology characteristics and Alternatives. Step 2 Data processing.

Grey data processing must be performed before GRGs can be calculated. A series of various units must be transformed to be dimensionless. This step is called 'grey relational generating'. The data can be treated with the following three situations. If the value of the original sequence is larger-the-better, it can be normalized by

$$X_{i}'(k) = \frac{x_{i}(k) - x_{i}^{\min}(k)}{x_{i}^{\min}(k) - x_{i}^{\min}(k)}$$
(1)

If the value of the original sequence is smaller-the-better, it can be normalized by

$$X'_{i}(k) = \frac{x_{i}^{\max}(k) - x_{i}(k)}{x_{i}^{\max}(k) - x_{i}^{\min}(k)}$$
(2)

But, if there is 'a specific target value', the original sequence is normalized using

$$X_{i}'(k) = 1 - \frac{x_{t} - x_{i}(k)}{x_{t}}$$
(3)

Where xt is the specific target value. After normalization, D becomes matrix D',

$$D = \begin{bmatrix} x_{11}' & \cdots & x_{m1}' \\ \vdots & \ddots & \vdots \\ x_{1n}' & \cdots & x_{mn}' \end{bmatrix}$$

Similarly, Y1 can be normalized as Y1',

.....

$$y_1 = (y_1'(1), y_1'(2), ..., y_1'(n))^T$$

Step 3 Calculations of grey relational coefficients. The grey relational coefficient is defined as follows (Dhas and Satheesh, 2012):

$$\gamma(\gamma_1'(k), x_i'(k)) = \frac{\hat{o}_{1\min} + \varepsilon \hat{o}_{1\max}}{\hat{o}_{1i}(k) + \varepsilon \hat{o}_{1\max}},\tag{4}$$

Where  $\delta 1i(k)$  is the deviation sequence of reference sequence y1'(k) and comparability sequence xi'(k),  $\delta 1i(k) = |y1'(k) - xi'(k)|$ ,  $\delta 1max = max max \delta 1i(k)$ ,  $\delta 1min = min min \delta 1i(k)$ ,  $\varepsilon$  is the distinguishing coefficient,  $\varepsilon \in [0, 1]$ .

Step 4 Determination of GRG. After the grey relational coefficients are got, the mean of the coefficients is often used as the GRG.

$$\gamma(Y_1, X_i) = \frac{1}{n} \sum_{k=1}^n \gamma(\gamma'_1(k), x'_i(K)).$$
(5)

Normalization is performed to get the final GRG.

$$\gamma(Y_1, X_i) = \gamma(Y_1, X_i) / \sum_{i=1}^{m} \gamma(Y_1, X_i)$$

(6)

Then  $\gamma$  (Y1, Xi) is the relation degree between the alternatives and attributes (Li *et al.*, 2008).

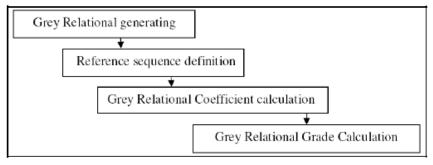


Fig. 2: Gray relational analysis.

Fig 3 shows the output of GRA method. And table 1 shows ranks and degree of confidence.

#### Feature selection:

Feature selection is one of the two important factors contributing to the performance of a prediction model for any classification problem. The objectives of feature selection are three-fold: (a) better performance, (b) faster and more cost-effective models, and (c). deeper insight into the underlying processes. (Guyon & Elisseeff, 2003). In many research problems, such as pattern recognition, it is important to choose a group of set of attributions with more prediction information. That is, if the number of irrelevant or redundant features is reduced drastically, the running time of a learning algorithm is also reduced. Moreover, a more general concept can be yielded. Performing feature selection can lead to many potential benefits, which are facilitating data visualization and data understanding, reducing the measurement and storage requirements, reducing training and utilization times, defying the curse of dimensionality to improve prediction performances, etc. [Guyon, Elisseeff, 2005] [Kim, 2006] [Mladenic´,Grobelnik, 2003].see feature selection process in fig 3 that run with clementine software.

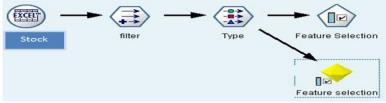
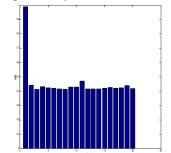


Fig. 3: Feature selection process by clementine software.



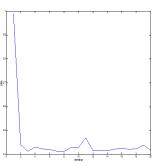


Fig. 3: Grey relational analysis to ranking of stocks.

1	Table 1: ranks and degree of confidence.										
	A5	A18	A16	A10	A9	A4	A19	A2	A11	A1	Stock
	0.4240	0.4246	0.4267	0.4293	0.4306	0.4316	0.4397	0.4411	0.4707	0.9898	GRA
	10	9	8	7	6	5	4	3	2	1	Rank
	A8	A3	A7	A14	A13	A12	A20	A6	A17	A15	Stock
	0.4130	0.4136	0.4148	0.4163	0.4165	0.4166	0.4184	0.4209	0.4218	0.4226	GRA
	20	19	18	17	16	15	14	13	12	11	Rank

Table 1: ranks and degree of confidence.

Conclusion:

The main aim of this research was selection optimized portfolio using grey relational analysis (GRA) and feature selection methods. First, this paper used of feature selection method to identify key indicators as criteria for recognition of proper firms. Second, uses of grey relational analysis (GRA) method to ranking of examined firms for selecting a basket of stocks that has the best performance. The Results of this study indicate that a grey relational analysis (GRA) technique is employed to specify an appropriate weighting of the selected stocks such that the portfolio rate of return is maximized. Finally, was found that, both of (GRA) and feature selection methods are fit. Performing feature selection can lead to many potential benefits, which are facilitating data visualization and data understanding, reducing the measurement and storage requirements, reducing training and utilization times, defying the curse of dimensionality to improve prediction performances, etc. in this paper used of this method to reduce indicators and lead to bounded group of indicators such as: company size, ROE, ROA, P/BV, stock price, Current Ratio.

## REFERENCES

Markowitz, H.M., 1952. Portfolio selection, J. Finance, 7: 77-91.

Speranza, M.G., 1996. A heuristic algorithm for a portfolio optimization model applied to the Milan Stock Market, Computers and Operations Research, 23: 433-441.

Mansini, R., M.G. Speranza, 1997. On selection a portfolio with fixed costs and minimum transaction lots, Report no. 134, Dip. Metodi Quantitativi, University of Brescia, Italy.

Mansini, R., M.G. Speranza, 1999. Heuristic algorithms for the portfolio selection problem with minimum transaction lots, European Journal of Operational Research, 114: 219-233.

Kellerer, H., R. Mansini, M.G. Speranza, 2000. Selecting portfolios with fixed costs and minimum transaction lots, Annals of Operations Research, 99: 287-304.

Konno, H., H. Yamazaki, 1991. Mean-absolute deviation portfolio optimization model and its application to Tokyo Stock Market, Management Science, 37: 519-531.

Konno, H., A. Wijayanayake, 2001. Portfolio optimization problem under concave transaction costs and minimal transaction unit constraints, Mathematical Programming 89: 233-250.

Mansini, R., M.G. Speranza, 2005. An exact approach for portfolio selection with transaction costs and rounds, IIE Transactions, 37: 919-929.

Deng, J.L., 1982. Control problems of grey system. Systems and Control letters, 1: 288-294.

Deng, J.L., 1988. Properties of relational space for grey system. In: Deng, J.L. (Ed.), Essential Topics on Grey System—Theory and Applications. China Ocean, Beijing, pp: 1-13.

Huang, S.J., N.H. Chiu, L.W., Chen, 2008. Integration of the Grey relational analysis with genetic algorithm for software effort estimation. European Journal of Operational Research, 188: 898-909.

Guyon, I., A. Elisseeff, 2003. An introduction to variable and feature selection, Journal of Machine Learning Research, 3: 1157-1182.

Kim, Y., 2006. Toward a successful CRM: variable selection, sampling, and ensemble, Decision Support Systems, 41(2): 542-553.

Mladenic, D., M. Grobelnik, 2003. Feature selection on hierarchy of web documents, Decision Support Systems, 35(1): 45-87.

Tzeng, G.H. and J.J. Huang, 2011. Multiple Attribute Decision Making: Methods and Applications, CRC Press, USA.

Li, Z., D. Zhang, and Gao, 2009. 'A grey method of prioritizing engineering characteristics in QFD', Paper presented at the Control and Decision Conference, pp: 3443-3447, IEEE2009, Chinese.

Kuo, Y., T. Yang, and G.W. Huang, 2008. 'The use of grey relational analysis in solving multiple attribute decision-making problems', Computers & Industrial Engineering, 55(1): 80-93.

Guyon, I., A. Elisseeff, 2003. An introduction to variable and feature selection. The Journal of Machine Learning Research, 3: 1157-1182.

Beshkooh, M., M.A. Afshari, 2012. Selection of the Optimal Portfolio Investment in Stock Market with a Hybrid Approach of Hierarchical Analysis (AHP) and Grey Theory Analysis (GRA).

Tsai, C.F., Y.C. Hsiao, 2010. Combining multiple feature selection methods for stock prediction: Union, intersection, and multi-intersection approaches. Decision Support Systems, 50(1): 258-269.

Lin, C. C., Y.T. Liu, 2008. Genetic algorithms for portfolio selection problems with minimum transaction lots. European Journal of Operational Research, 185(1): 393-404.

Wang, Y.F., 2003. Mining stock price using fuzzy rough set system. Expert Systems with Applications, 24(1): 13-23.

Huang, C.F., 2012. A hybrid stock selection model using genetic algorithms and support vector regression. Applied Soft Computing, 12(2): 807-818.

Bulgurcu, B.K. 2012. Application of TOPSIS Technique for Financial Performance Evaluation of Technology Firms in Istanbul Stock Exchange Market. Procedia-Social and Behavioral Sciences, 62: 1033-1040.

Hongyi, L.I., C. ZHANG, & Z.H.A.O. Di, 2011. Stock investment value analysis model based on AHP and gray relational degree. Management Science and Engineering, 4(4): 1-6.

Chen, Y., K. Hirasawa, 2011. A portfolio selection model using genetic relation algorithm and genetic network programming. IEEJ Transactions on Electrical and Electronic Engineering, 6(5): 403-413.