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# Particle Swarm Optimization Feature Selection for Data Stream Mining

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**ABSTRACT:** Big data concept is used in every industry or any business. Big data can be characterized by 3Vs: the extreme volume of data, the wide variety of types of data and the velocity at which the data must be processed. Although big data doesn't refer to any specific quantity, the term is often used when speaking about petabytes and Exabyte of data, much of which cannot be integrated easily. Big data contains not only the challenges that are coming from academic research communities but also commercial IT deployment, the root sources of Big Data are founded on data streams and the curse of dimensionality. It is generally known that data which are sourced from data streams accumulate continuously making traditional batch-based model induction algorithms infeasible for real-time data mining. Feature selection allows us to selection of data through its features we can filter the data using their features. But when the large dataset come the feature selection becomes complicated, and for the high dimension of data the feature which are used to fi9lter are also increased.

KEYWORDS: Particle Swarm Optimization, Swarm Intelligence, Classification, Big data.

## I. INTRODUCTION

People tend to invest in stock of a company due to its high rate of return. However, stock prices are continuously changing and most investors including experienced investors fail to understand the market trends. Investors are often influenced by crowd psychology and tend to go with majority decision of buying or selling stock which may be against the current market trend.

Hence, an effective financial system is required which can make objective decision of whether to buy or sell the stock. There are many prediction methods available for making stock prediction such as chartist method, regression analysis and neural networks. This project application uses the PSO for data optimization, feature selection and data mining for modeling the stock because of its ability to learn from the given training data and then generalize a rule.

The principal strength with the network is its ability to find patterns and irregularities as well as detecting multidimensional non-linear connections in data. The model is trained using the historical daily stock quotes of at-least two years. The historical stock quotes can be fed directly through the Yahoo Finance Server using internet or from the local database. Once the model is successfully trained, the model can be used to predict the value of share in future. This allows the investors to reduce the risk of investing, and enables easier decision making resulting in maximum financial gain<sup>[1]</sup>.

The brain is a multi-layer structure (think 6-7 layers of neurons, if we are talking about human cortex) with 10^11 neurons, structure, that works as a parallel computer capable of learning from the "feedback" it receives from the world and changing its design (think of the computer hardware changing while performing the task) by growing new neural links between neurons or altering activities of existing ones. To make picture a bit more complete, let's also mention, that a typical neuron is connected to 50-100 of the other neurons, sometimes, to itself, too.



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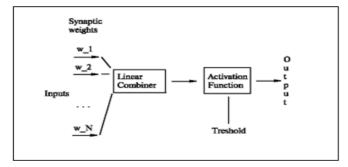


Fig 1: structure of Neuron.

This project mainly focuses on data optimization which will be given input to feature selection, and points out the shortage that exists in current traditional statistical analysis in the stock, then makes use of PSO for data optimization, feature trend selection and then uses Data mining algorithm to predict the stock market by establishing a three-tier structure of the neural network, namely input layer, hidden layer and output layer. After building the data pre-processing set before data mining, lots of widely used stock market technical indicators. Finally, we get a better predictive model to improve forecast accuracy.

In this project, we apply data mining technology to Indian stock market in order to research the trend of price, it aims to predict the future trend of the stock market and the fluctuation of price. Neural network is used to optimized the data, The Neural Networking algorithms (at least some of them) are modeled after the brain (not necessarily - human brain) and how it processes the information. The brain is a very efficient tool. Having about 100,000 times solver response time than computer chips, it (so far) beats the computer in complex tasks, such as image and sound recognition, motion control and so on. It is also about 10,000,000,000 times more efficient than the computer chip in terms of energy consumption per operation.

A neuron is an electrically excitable cell that processes and transmits information through electrical and chemical signals. Our "artificial" neuron will have inputs (all N of them) and one output:Set of nodes that connects it to inputs, output, or other neurons, these nodes are also called synapses.

Specifically, this paper makes the following contributions:

- Data optimization of the data which will be given input to the data mining.
- Till now Particle swarm optimization is exist only for the ECG system, this paper present PSO for the stock market dataset, so whatever data will be given for feature extraction is already optimized so that the accurate feature will be extracted and the prediction will be more accurate.

This paper continues to exist as follows: section II presents related work Section III is about design details of proposed PSO system. Finally the paper concludes in section V.

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The main purpose of energy efficient algorithm is to maximize the network lifetime. These algorithms are not just related to maximize the total energy consumption of the route but also to maximize the life time of each node in the network to increase the network lifetime. Energy efficient algorithms can be based on the two metrics: i) Minimizing total transmission energy ii) maximizing network lifetime. The first metric focuses on the total transmission energy used to send the packets from source to destination by selecting the large number of hops criteria. Second metric focuses on the residual batter energy level of entire network or individual battery energy of a node [1].



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#### II. RELATED WORK

## A. COLLABORATIVE FILTERING

Collaborative filtering: This method is based on the user behavior and prediction. The fundamental concept is that collect the opinion from the other users, in such a way to provide some prediction about the active user prediction. For example, if one person like item which is like by group of other personthen he will likely agreed upon the item which is like by that same group. Two types of Collaborative

• Memory based Collaborative filtering

Another name is lazy recommendation algorithm. It calculates the customer prediction for an objectwhen customers ask for collection of recommendation. Hence algorithms need to store the customerrating information in memory. There are two different algorithm based on k-nearest algorithm: 1.Itembased filtering, 2. Customer based filtering Item based filtering method is based on the most similaritems/objects. The item is considered for similarity when same group of customer ranked them highly.For every item which belong to a current customer, the neighborhood of same object is consideredand every top k-neighbor are put on the list, and recommendation are present to active user. Customerbased filtering compare the active customer/user to the other neighbor which have a past concurrence with the active user. First all neighbor are find out and then the entire object which belongs to neighborwhich are strange for active customer are regarded as a suggestion.

• Model based Collaborative Filtering:

This method is based on the dataset of rating, where we extract information from dataset, and use thatas a model to calculate recommendation with no use of entire dataset which benefits of both speedand scalability.

### **B. TUPLE BASED RECOMMENDATION SYSTEM**

Recommendation system used the query log registry which stored history of every old user who was use this system. And tuple based system used the data which is retrieved from previously executedquery. Every successfully executed query and there result are stored with the all output which containstuple data. The downside of this system is increase complexity. The size of data increases linearlywith increasing session summary. And every time we want to compare the current user session data with previous user stored data.

### C. FRAGMENT BASED RECOMMENDATION SYSTEM

This method implementation is same as the Tuple based recommendation method. The main difference is in method of session data comparison. In fragment base system, we use fragments of queryinstead of the witness means all tuple included data. In this method entered query is fragmented and those fragments are used to comparison. This method gives better result than tuple based recommendation.

### D. TRADITIONAL AND INCREMENTAL APPROACH

As the size of database increases the feature set also increases. Traditional system approach contains classification approach using top-down supervised learning Traditional classification approach, where a full set of data is used to construct a classification model, by recursively partitioning the datainto forming mapping relations for modelling a concept. The previous model is based on stationarydataset so whenever new item comes model update needs to repeat the whole training process. Sincethese models are built based on a stationary dataset, model update needs to repeat the whole training process whenever new samples arrive, adding them to incorporate the changing underlying patterns. The traditional models might have a good performance on a full set of historical data, and the dataare relatively stationary without anticipating much new changes. In dynamic stream processing environment, however, data streams are ever evolving and the classification model would have to befrequently updated accordingly. Therefore a new generation of algorithms, generally known as incremental classification algorithms or simply, data stream mining algorithms has been proposed to solvethis problem.



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# III. SYSTEM ARCHITECTURE

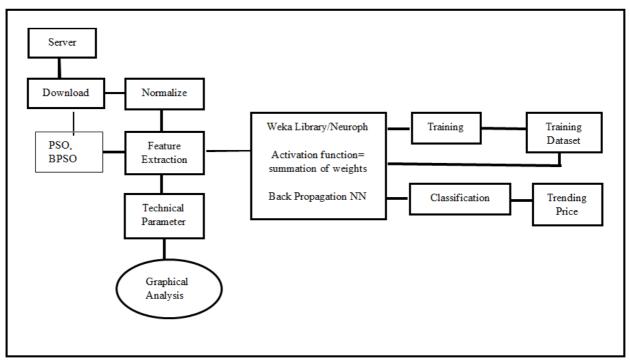


Fig.2. System Architecture

In this system the data input is downloaded from the server and given to the normalize module, the data is also given to the PSO and BPSO module which will optimize the data , the combination of both normalize data and the output of PSO and BPSO is given as a input to the feature extraction module. Technical parameters are nothing but the graphical comparison of data it will check the data by graph and the analysis or the differences are taken from the measure, so the feature extraction module are given to the back propagation algorithm which will contains the activation function which is summation of weights, the output is given to the training module which contains the patterns that are occurring against the time of period according to that the training dataset is created and given back to the module, then the classification is done for the calculating the price means the pattern is checked for the classification and then the trending price will be given to the end user.

### IV. PSEUDO CODE

### Algorithm 1: Backpropagation Algorithm

1) Assign all network inputs and output

- 2) Initialize all weights with small random numbers, typically between -1 and 1
- 3) Repeat
- 3.1) for every pattern in the training set
- 3.2) present the pattern to the network
- 3.3) for each layer in the network
- for every node in the layer
- 1. Calculate the weight sum of the inputs to the node
- 2. Add the threshold to the sum
- 3. Calculate the activation for the node
- End



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End

3.4) for every node in the output layer calculate the error signal End
3.5) for all hidden layers for every node in the layer
1. Calculate the node's signal error
2. Update each node's weight in the network
End
End
End
While ((maximum number of iterations i than specified) AND (Error Function than specified))
Algorithm 2: Calculating weight using neural network
1) Randomly choose the initial weights
2) INITIALIZE ERROR VAL

3) NORMALIZE STOCK VALUE STKVAL = STKVAL/MAX VAL

4) LOOP While error is too large (ERROR VAL = 0.5) For each training pattern (Set Input 4 Value and Set Desired

)

5) Apply the inputs to the network

6) Calculate the output for every neuron from the input layer, through the hidden layer(s), to the output layer

7) Calculate the error at the outputs

8) Use the output error to compute error signals for pre-output layers

9) Use the error signals to compute weight adjustments

10) Apply the weight adjustments

11) END LOOP

12) Periodically evaluate the network performance

### Algorithm 3: Particle swarm optimization

1) Initialize the particles with random velocities and positions in a given Dimension.

2) Compute the fitness of all particles using the desired benchmark function, choose the lowest one as global best and assign the current positions of all particles as their best.

3) Calculate the next velocities and positions using the equations 1 and 2.

4) Calculate the fitness of all particles using the updated velocities and positions. If the new fitness is less than the particles best fitness, the new fitness is considered the best one. In the same way the new position is considered the particle best position.

5) In the same way if the new particle's best fitness is less than the overall global best fitness, it is considered the new best global fitness and its corresponding position is considered new global best position.

6) Repeat third step for desired number of function evaluations.

# V. MATHEMATICAL MODEL

The proposed system S is defined as follows: S = I, O, F Where I: Input O: Output F: Functions U: User Where I= R, ND, OD Where R= Report of stock data ND= Normalized data



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OD= Optimized data

O = OD, FS, CD, TT, TPWhere OD = Optimized dataFS = Features Set.CD = Classified data.TT = Training tuple.TP = Trending Price.

F = PSO, BPSO, BP Where PSO = Particle swarm optimization BPSO = B-Positive Particle Swarm Optimization BP= Back Propagation

#### VI. SIMULATION RESULTS

#### System Implementation (Pre-processing)

The prototype of the proposed system is implemented using java.

#### Filter

The Data collected by the sensor is filtered using High Pass and Low Pass Filter. The filtered data is further given for Feature extraction Process

#### **Feature Extraction Normalization Process**

The Filtered data is normalized using the below featureextraction methods

I. Standard deviation of the NN The simplest variable to calculate is the SDNN that is the squareroot of variance. Since variance is mathematically equal to total power of spectral analysis, SDNNreflects all the cyclic components responsible for variability in the period of recording. In many studies, SDNN is calculated over a 24 hours period and thus encompasses both short-term high frequencyvariation, as well as the lowest frequency components seen in a 24-hours period, as the period ofmonitoring decreases, SDNN estimates shorter and shorter cycle lengths. It should also be noted thatthe total variance increases with the length of analyzed recording. Thus SDNN is not a well-defined statically quantity because of its dependence on the length of recording period. Thus, in practice, it isinappropriate to compare SDNN measures obtained from recordings of different durations.

$$s = \sqrt{\frac{1}{N-1} \sum_{i=1}^{N} (x_i - \overline{x})^2},$$

II. Standard deviation of differences between adjacent NN intervals. The most commonly used measures derived from interval differences include the standard deviation of differences between adjacent NN intervals. Calculation of standard deviation is show in above equation.

III. Root mean square successive difference of intervals the most commonly used measures derived from interval differences include the square root of the mean squared differences of successive NN intervals. Calculation of root mean square is show inequation.



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Fig 3 : Stock analysis based on time period

Fig 3, illustrate that when the authenticate user login to the system, he will able to see the different companies and their stock prices history, here user can select the duration such as last 1 month, last 6 month, according to that system will fetch the data and display it to user.

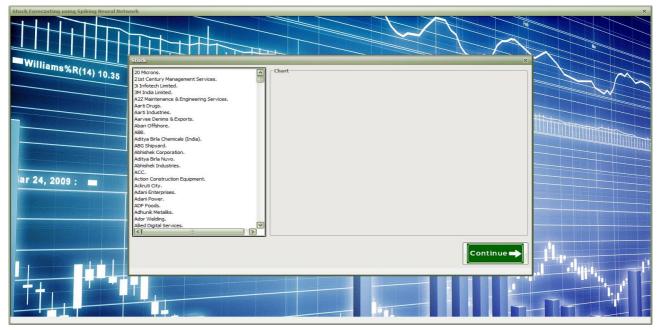


Fig 4 :Stock selection

Fig 4, describes the list of companies so that user will able to view the different companies and their past history till the date.



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Stock Code	SUZLON.NS						
⊙ Online [ ○ Offline	Download From	Date 7/1/2014	Browse	Get Quote			
Date	Open	High	Low	Close	Volume	Adjacent	
2016-07-01	17.45	17.75			16944500	17.20	1
2016-06-30	17.35	17.75 PS	0 Optimization	×	25266900	17.40	
2016-06-29	17.30	17.45			15277000	17.10	
2016-06-28	17.10	17.45	Dataset Size ha:	s reduced by 57 %	16216700	17.20	
2016-06-27	16.80	17.15			16945100	17.00	
2016-06-24	16.30	16.85		_	57967100	16.75	
2016-06-23	17.10	17.30	OK		22585900	17.05	
2016-06-22	17.50	17.60		-	17470400	17.15	
2016-06-21	17.65	17.80			18961300	17.65	
2016-06-20	17.50	17.90	17.35	17.45	34825000	17.45	
2016-06-17	18.00	18.25	17.45	17.65	33330500	17.65	
2016-06-16	18.10	18.15	17.10	17.80	56997300	17.80	
2016-06-15	18.15	18.45	18.05	18.15	46149400	18.15	
2016-06-14	17.50	18.00	17.35	17.90	54316700	17.90	
2016-06-13	17.00	17.30	16.85	16.95	29561600	16.95	
2016-06-10	17.20	17.70	16.90	17.25	55970500	17.25	I.

Fig5: Data reduction.

In Fig 3, we have selected "SUZLON" as a company, now in Fig 4, after selection of past duration for which I need to get quote, system will represent the data on daily basis, so each row contain open price, high price, low price, closing price for that day etc. after that I selected option "optimize" to optimize or to reduce data. Here I applied my algorithm Particle swarm optimization to the neural network (data). So we will get result, in this particular example, I have get result, data reduction by 57%.

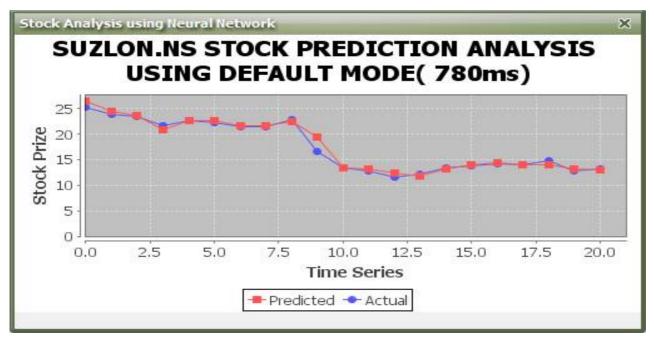


Fig 6 :Prediction without using PSO (Time difference).



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Fig 6, indicates the predicition of stock data using default mode. This will take 780 ms to analyze the data. These chart contains the two lines red line indicates the prediction of the stock and red line indicates actual price.

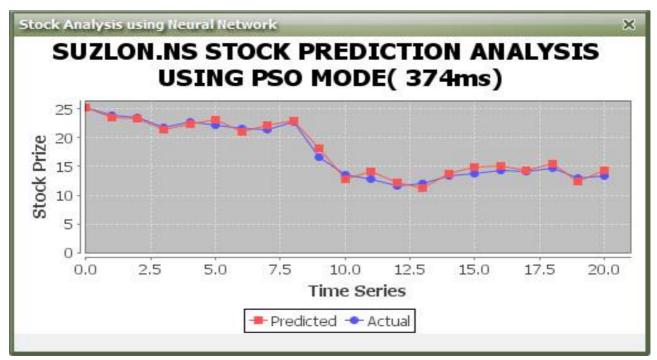


Fig 7 :Prediction with using PSO (Time difference).

Fig 7, contains information after applying particle swarm optimization algorithm to data. In fig 6 and fig 7 we can see the time differnce, without using PSO it is taking 780 ms and with using PSO it is taking 374ms which will reduce the time by 406ms. This indicates that PSO reduces the data and hence time required to calculate the analysis is also minimized. We can also see that after reduction also the result or data remains same.

A mutual fund is a pool of money from numerous investors who wish to save or make money just like you. Investing in a mutual fund can be a lot easier than buying and selling individual stocks and bonds on your own. Investors can sell their shares when they want.By investing in mutual funds, you could diversify your portfolio across a large number of securities so as to minimize risk. By spreading your money over numerous securities, which is what a mutual fund does, you need not worry about the fluctuation of the individual securities in the fund's portfolio.A closed-end fund has a fixed number of shares outstanding and operates for a fixed duration (generally ranging from 3 to 15 years). The fund would be open for subscription only during a specified period and there is an even balance of buyers and sellers, so someone would have to be selling in order for you to be able to buy it. Closed-end funds are also listed on the stock exchange so it is traded just like other stocks on an exchange or over the counter. Usually the redemption is also specified which means that they terminate on specified dates when the investors can redeem their units.An open-end fund is one that is available for subscription all through the year and is not listed on the stock exchanges. The majority of mutual funds are open-end fund's. Investors have the flexibility to buy or sell any part of their investment at any time at a price linked to the fund's Net Asset value.



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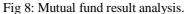


Fig 8, shows the analysis of data for multiple of companies, here I have combined three companies so AVG Analysis graph shows that the difference between each stock price and also last yellow bar is showing the average difference between 3 companies. Same for the percentage of change.

### VII. CONCLUSION

My present working is based on the optimizing the data that is feed to the prediction system and that will help people in analyzing the market condition. Challenging task is the amount of data daily feed into system is tremendous, and the share of different companies changes continuously so it need to be real time and the data must be updated according to time. There are different types of attribute which vary according to industry type so that parameter also need to be taken into consideration. The multi-objective PSO-based feature selection approach to better explore the Pareto front of non-dominated solutions in feature selection problems. To investigate the use of B-positive multi-objective PSO for feature selection and compare its performance with that of continuous multi-objective PSO.

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