Management Strategy for Mechanical Trade Systems: Brief Review

Musaev Alexander Azerovich*, Anantchenko Igor Viktorovitch and Gazul Stanislav Mikhailovich
Department of IT and Control systems, St. Petersburg State Institute of Technology, Russia

Abstract
Brief review and comparative analysis of management strategies, implemented in creating of mechanical trading systems (or trading robots) are presented in the article. A requirement for formation of effective trading solutions is the key feature of formation of such strategies in conditions of chaotic dynamics of trade assets quotations. Their nonstationarity and extremely low level of predictability are specific for the majority of processes observed in electronic exchanges.

Keywords: Mechanical trade systems; Trade robot; Management strategy; Chaos; Chaotic dynamics; Exchange market; Forex

Introduction
Studies in the field of management of open nonlinear systems in chaotic environments are one of the most relevant directions of modern science. The respective area of a study combined into a new scientific paradigm – synergetics – allowed to have a fresh look at the nature of uncertainty. In particular, until the middle of twentieth century random events and processes were described in accordance with probabilistically statistical conception. One of its promoters, Laplace (1749–1827), didn’t believe in accidents. It was him who derived the principle of determinism, according to which, all the processes are bound by strong cause effect relations. Absence of an unambiguous forecast was being explained solely by incompleteness of knowledge about all the aggregate of influences and factors of relationships between them. If a great know-all who received the name of “Laplace’s daemon” existed, he could just look and give an accurate forecast on the development for arbitrarily distant future without any difficulty. Only papers of twentieth century scientists - [1] and some others, who rethought the outstanding results of Poincare (1854-1912) in the field of nonlinear differential equations, included the definition of a natural source of uncertainty, as a parameter instability of nonlinear open dynamic systems in so-called bifurcation points. This finding led to complete change of notions about emergence of uncertainty and predetermined an emergence of new scientific paradigm of nonlinear interactions. The notions of turbulence, gas and hydrodynamics, specifics of information (non-material) processes, peculiarities of sociodynamics, etc. were substantially changed [2]. There was a new, highest stage of uncertainty described by chaotic processes [3,4].

The processes of change in quotations of currencies, stocks, raw material, and quotations of other assets in various capital markets became one of the vivid examples of chaotic dynamics. Hundreds of thousands of traders throughout the world are round the clock trying to forecast direction of currency instruments by time intervals in the range from several hours («scalping») to months and even years (the processes of strategic investment). More than 90% of traders are being ruined eventually. However, a few traders are still able to find weak links of order in chaos and steadily earn more than considerable capitals on their basis. This, in turn, means that the process of evolution of quotations is not pure chaos and holds local ordered segments in its structure. Two types of infrastructure and market data analysis are mainly used to detect these segments: Fundamental and technical.

Fundamental analysis (FA) uses available news and other information allowing to evaluate direction of movement of quotations in accordance with general economic, political and other considerations. Results of expert FA are regularly published on the broker and trader websites in the form of analytical reports and forecasts.

Technical analysis (TA) is related to formalized research of dynamics of quotations and uses all the arsenal of modern computer mathematics. In practice traders typically use a combination of both types of analysis.

Integrated informatization and automation [5] is an important trend of a present stage of exchange trade. The trading in electronic exchanges is being implemented globally through the internet network and automatic broker platforms [6]. Wide use of automated asset control systems and fully automatic systems – trading robots (TR) – became a powerful innovative trend.

Brief reviewing of modern state of management in chaotic environments and comparative analysis of basic approaches to construction of TR focused on functioning in the mentioned conditions is the topic of this article.

Trading Robots: Structure of Interaction and Main Objectives
The structure of interaction of TR with a broker platform is presented in Figure 1.

TR is a multifunctional automatic device [7]. They are assigned to carry out the following tasks:

1. Support of continuous telecommunication connections with a broker platform
2. Provision of online entries of current quotations, their pre-processing and detecting of system component
3. Implementation by TA of dynamics of quotations and formation of managing solutions D meeting a given game strategy
4. Formation of controlling signals and their transmission to a broker platform

TA is the key tool of formation of solutions for TR. However, advanced trade robots allow [8]:

- Consider limitations shaped by a trader in the course of FA

*Corresponding author: Musaev Alexander Azerovich, Department of IT and Control systems, St. Petersburg State Institute of Technology, Russia, Tel: +7(812)494-9323; E-mail: amusaev@technolog.edu.ru
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phenomenon was reviewed in detail in the monograph of Peters [13].

is the main feature of capital markets - their chaos. Corresponding
task, TR first of all needs to carry out the comparison of model trade
solutions derived from it [9,10]. Therefore, to resolve the comparison
entirely depend on the effectiveness of a management strategy and
important and decisive TR function. The results of trade speculations
of governing strategies.

It is obvious, that development of management solutions is the most
important and decisive TR function. The results of trade speculations
entirely depend on the effectiveness of a management strategy and
solutions derived from it [9,10]. Therefore, to resolve the comparison
task, TR first of all needs to carry out the comparison of model trade
strategies defining their functioning.

In conclusion we will formulate the list of requirements, which a
trade robot must meet:

1. TR must be a program application allowing obtaining sustainable
   receipt of profit from speculative operations in the financial
e-market
2. The TR program represents realization of an algorithm of a
   sustainable and profitable for a steward strategy of speculation
   in the speculative market [11]
3. The source data for the program are represented by series of
   observations on dynamic process of pricing in the financial
   market with strongly expressed chaotic structure
4. The program must allow to determine allowable level of loss
   being locked in SL preferably before opening an order for
   entering the position) and the level of profit TP fixation
5. Before the beginning of practical use, the trading system must be
   multiply tested in demo accounts and then, in case of positive
   testing results – cent accounts [12]

Stock exchange speculation: Basic definitions

The main problem for developing an effective management strategy
is the main feature of capital markets - their chaos. Corresponding
phenomenon was reviewed in detail in the monograph of Peters [13].

In particular, analysis presented in it showed, that capital markets,
contrary to abstract theories, were not effective and the quotations of
assets did not reflect their real value.

The process of quotation change is much informational (and
therefore, no inertial), subjective, ephemeral, dynamically unstable and
hyper open for an infinite number of influences [14]. In other words, it
corresponds to all the conditions of a chaos.

At present there are many definitions of chaos and chaotic
processes [15]. In this article we will understand chaotic process
as varying non periodic set of observations with an additive purely
random stationary component. Such a pattern is sufficiently typical for
quotation dynamics of market assets. Under the influence of numerous
information impacts, the quotation observation series create a dynamic
process of very complex nature almost not assuming formation of a
long-term or medium-term forecast [16,17].

A stochastic component which easily destroys traditional schemes
of statistics processing [18] by virtue of its nonstationary is present
here unlike conditioned (or dynamic) chaos. The key attribute of a
probability concept is absent - repeatability of experience.

Trade transaction management

The systems of trade transaction management in accordance with
their automation level can be divided into “manual” systems, where
decision is made by a person, and semi-automatic or automatic - trade
robots [19].

The semi-automatic systems or the counselors form, but do not
implement managerial decisions [20]. The final decision is made by the
trader who is comparing a counselors’ proposal to results of FA being
badly formalized, so the trader makes a final decision and shapes an
order for opening or closing the position.

Lastly, the trade robot does everything or almost everything on
its own. In some cases, the trader can implement human-machine
interface (HMI) of a robot, using some extra information (for example,
from FA), also setting necessary restrictions. For example, with
consideration for market conditions it can prohibit a robot to open a
position down during specified time interval [21].

The considered asset management options have their advantages
and disadvantages. But, in the opinion of analysts, a per cent of
trade transactions in electronic exchanges executed by robots is
constantly increasing. It is tied to both specificity of round-the-clock
dealings in the forex market and high requirements for promptness.
of solutions that are formed. Besides, the need of quantitative analysis of multidimensional and multi-concern information is basically out of reach for a human brain [22].

From the other side, the complexity of automatic analysis of badly formalized qualitative information and presence of so-called intuition (of those ones who have it), makes a trader's participation expedient for systems creating managing decision [23].

However, the paper considers only the issues related to trade robots [24]. The automated counsellors are using the same software as TR. The only difference is in form of representation of a managing decision.

**Market models**

Application of formalized managing strategies presupposes using a model or group of models reflecting quotation dynamics of market assets at formal level. Different options of same model development were considered [25,26].

In the opinion of the majority of specialists, it is extremely difficult to create a single model describing functioning of the financial market. In this connection, a group or a bank of models is created, the examples of which are:

- A model of growing market
- A model of stagnant market
- A model of volatile market
- A model of equilibrium market etc.

Models of growing and stagnant market typically have upward or downward trends. Such market can also be called 'trend' market ("bullish" or "bearish") it’s worth noting, that while estimating the trend and its force one should specify not only a trading instrument, but the time interval (time frame) used for analysis. Even if information of a junior timeframe (for example five-minute timeframe M5) is identified as a downward trend, it can look like a correction of an upward trend on older time frames [27-29].

In the model of volatile market the value of a price makes progressively-oscillatory motions with large amplitude, which allows opening diverse transactions in shorter or longer timeframes. This kind of trend movements are characterized by high correction shifts (kickbacks against trend), which should be considered while issuing stop orders.

Flat market prices (for considered financial instrument) almost don’t change, so there's basically no trend, there’s horizontal direction of movement (side trend or horizontal trend).

In practice, different types of the mentioned models can be used simultaneously. For example, a varying process with high level of volatility [30] can apply to the upward trend.

Figure 1 shows a chart for quotation dynamics of a currency pair of EURUSD during 150-day observation interval where the areas corresponding to different market models are clearly visible.

For any model of quotation change the task of building of an effective management strategy is trivial [31]. The problem is that it is impossible to guess what type of quotation dynamics should be expected in the nearest future, based only on trend analysis. The attempt of adaptation based on a prompt identification of dynamics type is also impossible to realize working with chaos [32]. The matter is that time, needed to identify a model is significantly higher than natural structural variability of structure of an observed process. Occurring lag leads to model inadequacy, and, as a result, inefficiency of a management strategy developed on its basis [33].

In conclusion, let’s note that a market model will significantly change depending on a type of operations conducted on it [34] (Figure 2).

For example, a model of conversion operations market or a model of the futures market. In a number of cases it is expedient to take into consideration territory-time peculiarities of market. In particular, the biggest session of the Asia-Pacific market is in the interval of 03:00-09:30 Moscow time, continent-European market – in the interval of 09:30-16:00, Great Britain market from 11:30 to 18:30, the US market – from 17:30 to 00:30. Division on sessions is sufficiently conditional and, as it was indicated, corresponds with the time of maximum activity in one or another region.

**Software (SW) for trade transaction management**

Classification of software for managing trade operations on capital markets can be viewed in accordance with different criteria:

1. Based on the purpose of software one is to distinguish programs for automatic (or automated) control and the programs supporting manual control. In particular, manual control is supported by indicator programs and script programs (or utilities) [35]

Indicators are the programs, which inform trader about the state of a trading instrument, current account or accounts parameters (account balance, resources, and the number of open orders, amount of non-fixed profit (or loss) on open orders, the information on the level of margin, etc. Scripts (utilities) are programs started by a trader, automating one of the operations (or example, to close all the orders of used financial instrument). The software for automated management includes above mentioned trade counselors and TR.

2. Based on type of software platform, used for trading (for example, MetaTrader 4, MetaTrader 5 are currently popular on Forex market)

3. Based on algorithms realizing the trading strategy

4. Based on additional functionality (for example, the possibility of sending messages to a mobile phone, e-mail, etc.) [36]

5. Based on the price of software

6. Based on the possibility of using client software in different operating systems

![Figure 2: Trend types at different observation stages of a currency pair quotation changes during 150-day observation interval.](Image)
7. Based on structural organization (single module, multi module, possibility of using dll libraries)
8. Based on the presence of additional tools to provide protection against unauthorized distribution and use (use of software or program-hardware protection tools) and licensing schemes [37]
9. Based on the complexity of trading program (processing of situations with temporary unavailability of trading server on a part of a customer, for example, for the reason of disconnection, etc.)

Development of effective programs for trade on financial markets becomes complex with a range of difficulties. We will note the most significant [38] of them:

The theory of algorithms, considering the specificity of management in chaotic environments is developed weakly at present. There are no algorithms allowing to consistently receive large profits for all possible states of market.

Software is developed on language of programming recommended for a used trading platform (for example, the MQL language version 4 for a MT4 terminal and MQL version 5 for a MT5 terminal). Programming language includes functions for working with trade orders, which allows executing specific trade transactions, which cannot be programmed on another language, not based on architecture of a used trading platform. On the other hand, a language may be insufficient for implementation of a complex algorithm of a program [39]. In this case one has to use DLL libraries written in other programming language. Development of such libraries and the correct function call (interface issues of program code of different development environments) often belong to a category of nontrivial and complex tasks for applied programming.

In the course of trade the program must consider a possibility of emergencies and support algorithms to process them. For example, if connection with a trade server was lost and the open orders lost support, then at the point of reconnection it may require conducting of the operations, which the program was not aimed at. The need for tracking and processing of additional conditions lead not only to expanding of a program code and time for its development, it decreases the speed of program [40].

Indicators – the software for supporting manual management of trade operations

It is important to note, that modern traders who don’t use automatic trading systems, don’t trade by the "bare" chart, they use different kinds of indicators. The indicators, as was mentioned above, represent formalized indicators, allowing quantifying the nature quotation dynamics of currency or some other financial instrument. Indicator is, first of all, a trader's instrument. But the ideas underlying in indicators are extremely useful in development of formalized trade strategies and trading robots [41].

Based on functionality we can divide indicators into 3 groups.

Indicators of a trend (trends) constitute the first group – they specify direction of price movements simultaneously or with a slight delay, and determine the continuation or the change of the trend. For example, trend indicators operating on estimation of volume change (accumulation/distribution, volume rate of change), trend indicators estimating change of moving averages of different periods (Bollinger bands, moving average), etc. [42].

Indicators-oscillators allowing predicting the continuation or the change of the trend in advance or almost simultaneously – constituting the second group. For example, Average True Range, Ichimoku Kinko Hyo, Detrended Price Oscillator, Price Rate of Change, etc.

Psychological indicators constitute the third group; they define sentiment of market participants. For example, there is a eToro informer displaying information about current purchase/sale deals, opened by 1000 best traders in the network of social trading eToro (http://www.mctraders.ru/files/etoro.zip) – visualization of information allows to evaluate a generalized opinion of leading experts about expediency of buying or selling of an examined financial instrument. If 90-99% of orders are opened for a purchase, the trend will most probably remain steady and it is expedient to consider a buying possibility, the same one refers to a situation of sale as well. If the ratio of buying and selling of orders is 50-80%, it testifies that leading traders do not have a common opinion in forecasts about change of these assets.

Management (trade) strategies

The key element of any TR is a program for managing decision forming, based on a specific (or managing) strategy [43, 44]. In this connection, fundamental differences between TR developing methods are the differences in managing strategies [45]. We will consider the most widespread options of managing strategies assuming a formalized representation, i.e. assuming its realization in the form of a program for developing managing decisions for trading robot. Let's consider formalized task for building a trading strategy.

\[ Y(t) \] - is the process of some asset quotation change. In the paper we use quotations of currency instruments on Forex e-market as an example. In order to decrease the influence of a purely random component, one can use different smoothing filters - exponential, Kalman filter, finite memory filter, moving averages, etc. Moving component \( Y(t) \) is some quazi system component, used for forming managing decisions \( D \). Here and further we talk about application of TA technologies, typical for building trading robots [46].

Note that electron nature of an exchange presupposes a digital form of observation sets \( Y(k), k = 1, ..., n \).

Managing strategy \( S \) is defined as a rule or a set of rules, providing functional representation \( S : \{ Y_1, ..., Y_{n_T} \} \rightarrow D_{n_T} \). Here \( T \) - depth of retrospective data used for present situation analysis, \( I \) is further information, for example, restrictions arising from a priori FA [47].

At its simplest, decision \( D_{n_T} \), consists of the following possibilities:
- Refrain from some actions;
- Enter a game i.e. to open the position "up" or "down";
- Close previously opened position.

Various options of complicated solutions are also possible. For example, managing decision can contain the levels of automatic take profit \( TP \) and stop loss \( SL \).

Let's consider particular types of managing strategies allowing implementation in TR [48,49].

Strategies on the basis of trend analysis

The main provision for building trend strategies is opening of a position in the direction of a discovered trend of quotation change of currency instrument used [50].
Basic trend strategy is based on the use of decisive rule (criterion), in accordance with which a position up is opened, if the growth speed of an observed process $a_1$, defined by slope of linear approximation on sliding window of observation of counts $w$ (for example, minutes, when using timeframe M1), is higher than a critical value $\alpha$. And on the contrary, position is opened under a condition $a_1 < -a_1$.

It is obvious, that one can generalize this strategy, using two or three trend analysis on different observation windows. The example of such an approach is a strategy based on the criterion $K$, using simultaneously two linear approximations on different sliding observation windows $w_1$ and $w_2$. A position up is opened under a condition of simultaneous excess of slopes in linear approximations of its critical values, i.e. $a_{11} > a_{11}^*$ and $a_{12} > a_{12}^*$. On the contrary, position down is opened, if $a_{11} < -a_{11}^*$ and $a_{12} < -a_{12}^*$.

The further development of trend strategies is linked to application of higher order trends. In particular, one can consider a strategy using a linear approximation on a sliding window $w_2$ simultaneously with a square approximation on a sliding window $w_c$. Position is opened up, under the condition $a_1 > a_1^* + a_2 > a_2^*$ or down, under condition $a_1 < -a_1^* + a_2 < -a_2^*$, where the parameters $a_1$, $a_2$ are essentially values of velocity and process acceleration on selected observation windows, and $a_1^*$, $a_2^*$.

For the first two strategies the estimates are of parameters of the slope of an approximating straight line $a_1$ or $a_{11}, a_{12}$, which can be computed on basis of least square method. However, if one uses polynomials in decisive statistics of higher degrees, the application of standard adaptation leads to a «mechanistic» approximation of least square method, where the estimates of coefficients of a polynomial cannot be interpreted as speed or acceleration of a process that is explored. This, in turn, deprives the possibility of process visualization while displaying trends graphically.

The trend strategies are beneficial in areas with clearly expressed trends to growth of quotations or quotation drop. But in the areas with the flat trend, they, as a rule, lead to a significant loss [51,52].

### Strategies on the basis of oscillators

This class of managing strategies is based on use of so-called oscillators, which are the estimates of propriety of current market value of a financial instrument. It is considered that the oscillators are indicators of «overbought» and «oversold» state of financial instrument. The examples of oscillators suggested for Forex are «Stochastics», «Relative strength index (RSI)», «Stochastic Expansion», etc.

A computational scheme on basis of dynamic limits of quotation observations variations $Y(k), k = 1,...,n$ can serve as the simple example of building a managing strategy on the basis of oscillating indicator. As a result of smoothing with a low transmission coefficient one can construct a process imitating the value of a moving average $Y_{1}(k) = aY(k) + (1 - a)Y_{1}(k-1), k = 2,...,n$. Essentially one can use a moving average procedure itself, or even other smoothing filter. The threshold values are set at the level of estimates of mean square deviation of a process $b = s(\Delta(k), k = 1,...,n) = \sqrt{\sum_{i=1}^{n} (\Delta^2(k))/ (m - 1)}$.

For optimization of the process of settlement one can vary threshold size, for example, in the range of $[0.75-2]$. Note that the process is stationary and estimate of mean square deviation is incorrect. Nevertheless, it is quite suitable for practical use in the first approximation.

The further decision is obvious. The evaluation criterion is that the quotation comes to the dynamic limit $Y(k) > Y_{1}(k) + b$, respectively, underestimation of an instrument is determined from a criterion $Y(k) < Y_{1}(k) - b$. In the first case the robot opens position down, in the second case – up.

The proposed strategy can be quite effective at felt stations and clearly losing when there is an expressed trend in dynamics of quotations.

More complex ways of incorrect estimation of a currency instrument indication are related to the use of correlational and regression analysis technique. Correlational analysis allows specifying a stochastic linear relationship between different financial instruments. At the same time, one can consider not only intramarket instruments (for example, currency pairs in the forex), but intermarket relationships as well. For example, it is easy to set a relationship between the quotation of a currency pair of EURUSD and quotations of DJ index of a New York Stock Exchange or quotations of oil of a goods-raw material exchange.

In a case, if regressive evaluation of quotation turns out higher than the value of a current quotation and their difference exceeds a set threshold value $\hat{Y}(k) - Y(k)$, it means that a financial instrument’s being undervalued and indicates a possibility of opening a position up. And on the contrary, the condition can $\hat{Y}(k) - Y(k) < -b$ serve as a recommendation for opening a position down.

It is obvious, that in conditions of strong trends this strategy can turn out ineffective. Besides, the situation is possible when not an instrument, but a regressor itself [53,54] is incorrectly estimated. In this case it makes sense to verify a result on another instrument correlated with a working quotation or, to carry out assessment by the group of regressors using multiregressional analysis.

### Precedent strategies

The precedent managing strategies (or strategies based on use of «patterns») are sufficiently typical even for manual asset control. In particular, many traders are trying to find standard patterns in quotation charts, which preceded standard changes of quotations. In particular, such patterns are well known - «wedge», «the pennant», «the flag», «the head are shoulders» etc. As a rule, traders underestimate cunning of «patterns») are sufficiently typical even for manual asset control. In particular, many traders are trying to find standard patterns in quotation charts, which preceded standard changes of quotations. In particular, such patterns are well known - «wedge», «the pennant», «the flag», «the head are shoulders», etc. As a rule, traders underestimate cunning of «patterns») are sufficiently typical even for manual asset control. In particular, many traders are trying to find standard patterns in quotation charts, which preceded standard changes of quotations. In particular, such patterns are well known - «wedge», «the pennant», «the flag», «the head are shoulders», etc. As a rule, traders underestimate cunning of «patterns») are sufficiently typical even for manual asset control. In particular, many traders are trying to find standard patterns in quotation charts, which preceded standard changes of quotations. In particular, such patterns are well known - «wedge», «the pennant», «the flag», «the head are shoulders», etc. As a rule, traders underestimate cunning of «patterns») are sufficiently typical even for manual asset control. In particular, many traders are trying to find standard patterns in quotation charts, which preceded standard changes of quotations. In particular, such patterns are well known - «wedge», «the pennant», «the flag», «the head are shoulders», etc. As a rule, traders underestimate cunning of «patterns», etc. As a rule, traders underestimate cunning of «patterns», etc. As a rule, traders underestimate cunning of «patterns», etc. As a rule, traders underestimate cunning of «patterns», etc.

To enhance veracity of precedent decision, dynamics of quotations must be reviewed in the large (several years) observation interval and, using non-parameter measures of similarity, estimate probabilities of stable after-effects. To estimate the extent of similarity one can use different statistical measures of proximity: $T^2_{0}$ Hotelling, Kolmogorov-Smirnov, $\alpha^2$ Mises, Wilkes, Kulbska-Laiblera, the traces of Pillai et al.
Market-neutral strategies

The market-neutral strategies (MNS) are strategies of revenue and risk which do not depend on market direction.

Paired trading is a variety of MNS being applied in the financial markets. A concept of cointegration developed by Engle and Granger [38] is theoretical basis of paired trading. For explanation of its nature we will introduce a number of definitions.

The ephemeral time series, which becomes stationary upon $n$ time differentiation becomes stationary is called integrated of order $n$ and is denoted as $I(n)$. Two integrated time series are $x$ and $y$ co-integrated if a linear combination of these series $z=ax+by$ exists and makes a stationary process. If a certain linear combination of two time series has a smaller procedure of integration than a characteristic of integration of each of the series, it is said that the time series are co-integrated.

The paired trading strategy allows to profit on a short-term imbalance in yield or prices on assets with high degree of correlation. One and the same external factors have impact on a couple of similar companies from one sector of economy. Therefore, the share prices usually set $(TP=SL)$.

Nevertheless, paired trading has a risk that the imbalance between assets keeps a trend character, i.e. a correlation between instruments becomes broken. Study of an issue of paired relationships in the forex market can be found [57].

Martingale strategies

Martingale strategies were initially used in gambling. The principle of a strategy is about successive lot growth by 40–100% in case of a loss. At the same time equal values of positive and negative result of play are usually set $(TP=SL)$.

The strategy was firstly proposed by the French mathematician Paul Pierre Levy and arose from multiplication rules of probabilities for independent events. One can come to the same strategy proceeding from Bernoulli’s theorem.

The strategy name originates from a French word of martingale denoting a part of horse harness – a semi-garrote limiting the position of a horse’s head. If there is a wrong move, the martingale slowly tightens and doesn’t let the horse to raise head. The similar device is used in the tackle of sail ships.

The disadvantages of such approach are sufficiently evident. The strategy requires a sufficiently large deposit and is extremely risky. In the opinion of experts, application of a martingale strategy allows to increase the possibility of profitable deals to 87% even in operation with a minimum deposit (4 financial margins). However, the risk is very high, around 62% and the deposit may be lost in one transaction sequence [57].

Scalping strategies

This approach, strictly speaking, is not a strategy, understanding under this term a set of rules unambiguously defining conditions of entering and leaving the market. This is some game ideological which one can describe more simply as an attempt for one to get small, but frequent profit with low risks [58,59].

As a rule, a scalping trader is working with short timeframes M1-M5 and tries to find a trend. After opening a position and receiving a minimum profit the position closes. Work is carried out within the day during which position can be taken 30–1000 times. At the same time the USD revenue for an operation of $(0.1–0.2)$ USD with risk $(0.02–0.03)$ can well suit the trader. In other words, the trader enters the market, scalped it and quickly left until the market punished him/her for presumption.

Many experts consider speculation ideal technology for trading training; on the contrary, the others are sure it requires great experience. Some examples of day trade are described in Shimlovići et al. [17].

It is obvious; that such a game is possible only with small amounts of commission (‘spreads’) established by a broker. In practice, it must not exceed 3–4 pips. A large volume of lot is required for successful operation. In this connection they try to use a large leverage often of 1:200–1:500 [41,60].

The scalping work is extremely redious and nervous. It can quite turn rational for the strategies of such a class to start using trade robots.

Conclusion

Despite a sufficiently large number of developed programs for trade in the financial markets, including Forex, one can make conclude that the vast majority of them are not universal and can bring profit only under certain conditions [61]. At the same time it is extremely difficult to determine observation intervals meeting conditions of effectiveness of these or other strategies. The delays related to the identification process do not allow switching to the required type of managing strategy in time, which can lead to significant losses, even the complete loss of a deposit. Nevertheless, the development of program-algorithmic software of trading robots realizing various variants of managing strategies represents not only scientific but also quite particular commercial interest. For example, on the portal of trading platform MT4 MQL4 Code Base (http://codebase.mql4.com/ru/) on 30.01.2014 93865 participants were registered and able to load 3219 programs (scripts) for trade in the Forex market, there are free, as well as commercial programs among them.

On another side the task of building effective managing strategies for trading robots is an applied task of a fundamental scientific issue of forecasting and management in chaotic environments. Even the partial solution of this problem, if it is ever made, discovers the broadest spectrum of practical applications in various fields of science, technology and socio-economic sphere.

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