Exploration of Stock Predictive Analysis Indicators and Innovative Forecasting Models

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Abstract: Stock price forecasting is the introduction of future stock prices based on the knowledge of past and present stock price data, which determines that the subject of stock forecasting research is not a specific event, but a random and uncertain event which requires consistent scientific forecasting methods to make a reasonable prognosis of investment. The purpose of this paper is to discuss the analytical indicators of stock forecasting and the application of innovative predictive models in stock market forecasting. Stock market forecasting is done by analyzing various factors such as past stock prices, financial data, market sentiment, and by using advanced techniques such as artificial intelligence, time series, and ensemble learning.

Keywords: stocks; predictive indicators; forecasting models

1. Introduction

The stock market is a very complex place to invest. For investors, it is of utmost importance to explore the inner laws of the stock market, look for effective analytical methods and tools to predict the direction of the stock market and predict the price of stocks. In practice, there are many factors that affect the stock market, which naturally leads to an increase in the investment risk. Predicting stock prices is a challenging problem because the volatility and non-linear nature of the stock market make it difficult to predict accurately. In recent years, breakthroughs in stock price forecasting have been achieved by a number of new methods that utilize external knowledge to better understand the stock market. However, a comprehensive summary of these methods is still insufficient.

Stock prices are affected by a variety of factors, including macro factors such as politics and culture, industry and region, company and market, thus creating complexity and instability in stock price forecasting. The stock market has rich data resources, with a lot of hidden information to be uncovered. Data mining means uncovering some of the potential rules and laws that seem to be contrary to common sense from a variety of massive data and use them to predict what the future will happen and make the corresponding decision. In this paper, we systematically introduce how to obtain external knowledge from various unstructured data sources and incorporate it into stock price prediction models, and innovate stock prediction methods to improve the accuracy of stock price prediction.

2. Research and Content of Stock Forecasting Models

Stock forecasting models can predict future stock prices by analyzing and modeling past stock market data. Their research mainly includes data collection and processing, feature engineering, model selection and evaluation.[1]

2.1 Data Collection and Processing

Data collection is the first step in stock forecasting models, which requires the collection of a large amount of stock market data, including stock prices, trading volume, financial data, etc. These data can be obtained from stock exchanges, financial data service providers and other sources. These data then need to be cleaned and processed to remove outliers and missing values to ensure the accuracy and completeness of the data.[2]

2.2 Feature Engineering

Feature engineering is a key aspect of stock forecasting models, which aims to extract and build
meaningful features from raw data for model learning and prediction. Commonly used features include technical indicators (e.g., moving averages, relative strength index, etc.), fundamental indicators (e.g., profitability, solvency, etc.), and market sentiment indicators (e.g., news sentiment analysis, public opinion heat, etc.). The quality and selection of feature engineering has a significant impact on the performance and accuracy of the model.

2.3 Model Selection and Evaluation

Model selection is to choose a suitable model for stock prediction from a wide range of machine learning algorithms. Commonly used models include linear regression, support vector machines, decision trees, neural networks, and so on. After selecting the model, it needs to be trained and evaluated. Evaluation metrics include mean square error, accuracy, return, etc., which can be derived through methods such as cross-validation and back-testing.

3. Key Analytical Indicators for Stock Forecasting

3.1 Financial Indicators

Financial indicators are important indicators that reflect the financial status of a company, including revenue, profit, assets and liabilities. By analyzing the financial indicators, we can understand the profitability, solvency and operating conditions of the company and provide a basis for stock forecasting. First, the revenue growth rate reflecting the changes in the increase or decrease of the company’s revenue, is an important indicator for assessing the company’s growth and profitability. Second, the net profit ratio, reflecting the proportion of the company’s net profit to operating income, is an important indicator for assessing the company’s profitability. Third, gearing ratio, reflecting the proportion of the company’s total liabilities to total assets, is an important indicator for assessing the company’s solvency and risk profile. Fourth, earnings per share, reflecting the company’s profitability per share of stock, is an important indicator for assessing the valuation of the company’s stock. Fifth, price-earnings ratio, reflecting the valuation level of the company’s stock, is an important indicator for assessing the investment value of the company’s stock.

3.2 Macroeconomic Indicators

Figure 1: GDP data of China(data source: Wind database)

Macroeconomic indicators are important indicators that reflect the state of national economic development, including GDP growth rate, inflation rate, interest rates, etc. By analyzing macroeconomic indicators, we can understand the national economic development trend and policy direction, which provides the basis for stock forecasting. They mainly include GDP growth rate, which reflects the country’s economic growth rate and is an important indicator for assessing the country’s economic development status. And usually, a high GDP growth rate is associated with stock market rise because
economic growth means that the company’s profit may increase (as shown in figure 1). Besides, Interest rates are also an important indicator. Interest rate reflects the country’s cost of funds and supply and demand in the financial market, and it is an important indicator for assessing the country’s monetary policy and the state of the financial market.

3.3 Company’s Fundamental Indicators

Company’s fundamental indicators are important indicators that reflect the company’s landscape, including corporate governance structure, development strategy, market competitiveness and so on. By analyzing the company’s fundamental indicators, we can understand the company’s competitiveness and development potential, which provides a basis for stock prediction. The first is the corporate governance structure, an important indicator for assessing the company’s operational efficiency and risk control, which reflects the company’s internal management and supervision mechanism. The second is development strategy, an important indicator for assessing the company’s future development and competitiveness, reflecting the company’s future development goals and plans. The third is market competitiveness, an important indicator for assessing the company’s future development and profit potential, reflecting the company’s competitive position and market share in the market.[3]

3.4 Market Sentiment Indicators

Market sentiment indicators are important indicators that reflect the mentality and psychology of investors, including investor confidence index and market volatility. By analyzing market sentiment indicators, we can understand the psychology and mentality changes of investors, which provides reference for stock forecasting. On the one hand, the investor confidence index, an important indicator for assessing market sentiment and future trends, reflects investors’ confidence and expectations of the market. On the other hand, the market volatility, an important indicator for assessing market sentiment and risk profile, reflects the magnitude of market price fluctuations and the level of risk.

3.5 Technical Analysis Indicators

Technical analysis indicators are indicators derived from mathematical calculations and statistical analysis based on changes in stock prices and volume. Through the analysis of technical analysis indicators, we can understand the trend of the stock, which provides reference for stock forecasting. Commonly used technical analysis indicators include moving averages, relative strength index (RSI), Bollinger Bands, etc. Moving averages, important indicators for evaluating stock trends and buy/sell signals, reflect the average level of stock prices over a certain period of time. Relative strength index (RSI), also an important indicator for evaluating the strength of a stock and buy/sell signals, reflects the overbought and oversold situation of a stock. Bollinger Bands combine standard deviation and moving averages to calculate the upper and lower limits of the normal distribution to show the distribution of fluctuations and the upper and lower limits, as a reference for buying and selling.

4. An Innovative Exploration of Stock Forecasting Models

4.1 AI-based Forecasting Models

With the increasing development of artificial intelligence (AI) technology, more and more researchers are trying to apply artificial intelligence to stock prediction. Among them, deep learning models are most widely used. With deep learning models, useful features can be effectively learned from past data and utilized for stock prediction. In addition, some researchers have attempted to use artificial intelligence techniques such as reinforcement learning and genetic algorithms to optimize stock trading strategies for higher investment returns.[4]

4.2 Time Series-based Forecasting Models

Time series forecasting modeling is a forecasting method based on time series data. In stock forecasting, this method is often used to predict stock price trends. By analyzing the time series of past price data, information such as autocorrelation and seasonal patterns of stock prices can be extracted and utilized for stock price forecasting. Some researchers have also tried to use deep learning models such as recurrent neural networks to improve the accuracy of time series forecasting and improve time series
anomaly detection for stock market forecasting. Quickly identifying underperforming stocks from thousands of stocks is a practical problem in unstable financial markets. Common stock market forecasting models fail to take outliers into account, so time for anomaly detection is needed to capture optimal trading points and reduce investment risks. The model can also be used for modeling and anomaly detection for multiple financial time series datasets. Therefore, designing a better mechanism based on time series anomaly detection task is an important future research direction in order to capture the best trading points for forecasting tasks in real world trading (as shown in figure 2).[5]

Figure 2: Time series documenting long-term stock changes

4.3 Ensemble Learning-based Forecasting Models

Ensemble learning is a method to improve prediction accuracy by integrating several different models. In stock prediction, ensemble learning can be used to integrate several different prediction models such as linear regression, decision trees, neural networks, etc. By integrating the predictions from multiple models, more accurate predictions than a single model can be obtained. Some researchers have also attempted to use optimization methods such as genetic algorithms to select the optimal combination of models for higher prediction accuracy. In addition, deep learning predictive modeling is one of the methods for stock prediction through deep learning techniques. By constructing deep neural network models, a large amount of past data can be automatically learned and optimized to gain the future movements and trends of stocks. Deep learning technology has powerful feature learning and processing capabilities, which can better handle complex nonlinear relationships, thus improving the accuracy and stability of prediction. Stock markets are characterized by uncertainty and variability, which makes it challenging to accurately predict market trends. And machine learning techniques have been used in stock price forecasting to improve the accuracy of predictions and alleviate these difficulties. Historically, traditional models such as decision tree-based models and support vector machines (SVMs) have been used for stock market forecasting.[6]

4.4 Scenario Analysis-based Forecasting Models

Scenario analysis is a method of forecasting based on possible future scenarios. In stock forecasting, scenario analysis can be used to predict different possible future market scenarios and the movements of stock prices under each scenario. By considering different market scenarios, the risks and returns of stock prices can be assessed more comprehensively, leading to a more robust investment strategy. Some researchers have also attempted to model the stochastic process of stock prices by statistical methods such as Monte Carlo simulation for more accurate scenario analysis.[7]

4.5 Social Media-based Forecasting Models

Social media is one of the most important sources of reflecting market sentiment. By analyzing the statements and emotions in social media, it is possible to understand the psychology and expectations of market participants, which can have an impact on stock prices. Some researchers have tried to use techniques such as natural language processing to extract sentiment information in social media and use it as one of the important features for predicting stock prices. This approach can help us better understand the impact of market sentiment on stock prices and develop more accurate investment strategies.[8]
5. Conclusions and Prospects

This paper demonstrates the importance of multiple factors in stock market forecasting and the effectiveness of advanced techniques in stock forecasting by exploring analytical metrics and innovative forecasting models for stock forecasting; in addition, we provide an overview of the main current methods, evaluation metrics, and datasets used for stock market forecasting. Through a multifaceted analysis, the paper demonstrates the effectiveness of the practical application of innovative forecasting models in stock market forecasting. In the future, with the continuous advancement of technology and changes in the market environment, further research on more accurate forecasting models and techniques is needed to better serve investment decisions.

References


