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Research on Theoretical Analysis and Optimization Path of Strategic Human Resource Management——Based on the perspective of Labor Economics

Wang Shuang [1*]

[1] School of Economics and Management, Guangxi Normal University, Guilin 541006, China
*Corresponding author Email: m15678306867@163.com
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Abstract: With the deepening development of economic globalization and the increasing intensity of social competition, the striving for and cultivation of human resources by economic entities has gradually become one of the inevitable choices for their survival and development, and the human resource management system that matches the needs of social development and the strategic objectives of enterprises has gradually become an important competitive advantage of enterprises. Therefore, each enterprise must take its own ability endowment as the foundation, fully understand the problems and needs of social and economic development, and use the high-quality solutions created by human resources as the tool to highly match the development trend of society, so as to promote the realization of the enterprise's economic and social benefits. Based on this, this paper discusses the application value of labor economics in strategic human resource management, introduces the basic content of strategic human resource management, and puts forward the effective landing strategy of strategic human resource management based on the perspective of labor economics.

Keywords: Strategic Human Resource Management; Labor Economics; Theoretical Analysis; Optimization Paths

1. Introduction

At the current stage, human resources have become the first strategic element to promote social development, playing an important role in supporting economic growth and promoting high-quality development. Many enterprises have also recognized the importance of human resource and begun to pay attention to the talent work. In this case, the concept, objectives, and models of human resource management of enterprises have undergone significant changes, gradually evolving from the previous personnel and professional function management model to the strategic human resource management model [1]. The transformation of management objective from simple "control of labor costs and support on the business operation of other units" to the "highlight of the strategic position of talent leading development is also happening [2], so the effectiveness of human resources can be maximized and the

enterprise strategy can be guided". Looking at human resources management from the strategic level and integrating human resources management activities with mission, vision and strategic planning are the trend of enterprise practice in recent years, and also an important connotation of strategic human resources management.

Strategic human resource management has been regarded as an important resource base for enterprises to gain performance advantages and realize sustainable development [3]. Existing theoretical studies have emphasized the importance of strategic human resources management from the construct and contingency views [4], with the construct view emphasizing the fit between the internal components of human resources management from the internal consistency view [5], and the contingency view considering the dynamic capabilities and resource base as an important path for strategic human resources management to influence performance [6]. In addition, strategic human resources management is strongly oriented to corporate goals, and "matching" provides a way to realize the success of human resources management in serving corporate goals, thus becoming an important perspective in strategic human resources management research [7]. Based on the matching view, scholars have identified the objects that human resources management needs to match (e.g., leadership [8], external environment [9]), and strategic human resources management integrates organizational and human resource strategies, enabling human resources management to understand the company's strategy and mission and to maintain consistent dynamics with them, which provides a wide range of technical support and organizational capability support for the achievement of strategic goals. By continuously adding value to human resources, it will make the company a more accessible and profitable organization and ultimately make it more competitive in a competitive environment. Therefore, it is important for companies to analyze the supply and demand of human resources based on labor economics to provide a basis for strategic human resource management.

2. The overview of Labor Economics

2.1. The Concept of Labor Economics

The essence of the proposal of labor economics is actually the scientific and effective combination of labor and the economic benefits generated through labor, because in economic activities, labor is its vital component, and it can also be said that without labor it is difficult to generate economic benefits. Labor as a special commodity, if its value can't be really exerted, it will affect the economic benefits to a great extent. At present, it is necessary to scientifically input the various elements of labor into the economic value, and through the labor supply needs to carry out an in-depth analysis to achieve the balance of the labor market, so as to produce the maximum economic benefits with the minimum labor input.

2.2. The application value of labor economics in strategic human resource management

In the current human resource management work, strategic human resource is an important task that needs to be concerned about, and requires that in human resource management must pay high

attention to the overall development strategy of the enterprise, and on this basis, human resource management is constantly optimized and adjusted, so as to make it more compatible with the development of the enterprise, to achieve the effective allocation of positions and improve the existing organizational structure, so as to achieve a win-win situation for both employees and enterprises. From the implementation of strategic human resources, the introduction of labor economics is very necessary, and its value is mainly reflected in:

On the one hand, it can effectively control the cost of human resource management. To a large extent, the application of labor economics in human resource management can effectively control the cost of human resource management, not only to save the expenditure of management funds, but also to avoid the problem of human cost waste. Strategic human resource management requires minimum input to obtain maximum output in management, and constantly optimize and improve the allocation of human resources, and if there is labor economics as a guide, it can help enterprises to effectively solve the situation of excess human resources, better improve the rational allocation of human resources, and improve the efficiency of management. At the same time, it can also provide an effective guarantee for enterprises to obtain more economic benefits.

On the other hand, improve the work efficiency. The good use of labor economics to human resource management work can maximize the efficiency of the staff to achieve the optimization of human resource management. There are many enterprises in the allocation of human resources are unreasonable, or in the talent incentives a great lack of these problems will largely affect the staff's work efficiency and effectiveness, so this is an issue that must be addressed in current human resource management. The effective application of labor economics to strategic human resource management can effectively remedy these problems and deficiencies. Analyzing the demand for human resources from the perspective of enterprise strategy not only improves the efficiency of human resource management, but also ensures the realization of the strategic objectives of the enterprise and gives full play to the maximum value and significance of strategic human resource management.

3. Theoretical analysis of strategic human resource management

Strategic Human Resource Management (SHRM) emerged in the 1980s and has been widely discussed and practiced in European, American, and Japanese enterprises. SHRM refers to a series of planned deployment and management behaviors of human resources for the purpose of enabling the enterprise to achieve its goals. It encompasses both the human resource activities that influence the behavior and commitment of individual enterprises in the process of formulating and implementing the strategic needs of the enterprise, as well as the integration of human resources at different functional levels to achieve the objectives of the enterprise. Strategic human resource management puts human resource management at a strategic height and focuses on the promotion of human resource management to the realization of organizational goals, which is a major breakthrough and continuation of the development of the traditional "simple personnel management ideas” and has been proved to be an important way for enterprises to obtain long-term sustainable competitive advantage. Compared with traditional human resource management, it has significant economic value.

3.1. Basic concepts of strategic human resources management

Strategic human resource management refers to the enterprise management based on strategic development goals from top to bottom of the human resources positioning, operation, evaluation, optimization, and improvement. It takes value creation and transmission as the core to synchronize human resources for maximum frequency collaboration and realizes the maximization of the enterprise strategic direction and path with the human resources management concept, configuration structure, compensation model, performance system and training system. Strategic human resource management requires enterprises to have a full understanding of the development needs of the social market and a deep insight into the organizational resources they need to complete the landing of the solutions. Thus, the action path, personnel scale, standard parameters and target results for human resources to realize the economic and social value of these resources are quantitatively, executably and evaluably designed, and the allocation of human resources is front-loaded rather than back-loaded to ultimately form a dynamic human resources system capable of responding, tracking, feeding back and optimizing and improving in accordance with the objectives in a timely and speedy manner [10].

3.2. The basic characteristics of strategic human resource management

Strategic human resource management is a resource allocation system that matches the overall development strategy of the enterprise, which is based on the contribution of human resources to the survival and development of the enterprise, and the symbiotic relationship with the enterprise's business system, operation system, production system, supply system, and financial system, etc., which are equal and collaborative. Horizontally, top-down human resources configuration structure design, and even any other functional modules of the enterprise must be based on human resources allocation; vertically, the value of the enterprise's resources to explore and create, market demand insight and satisfaction, internal transaction costs and reduce the balance of organizational costs and profits cannot be separated from the scientific and leveraged design of human resources.

3.3. The difference between strategic human resource management and traditional personnel management

Modernized strategic human resource management and traditional personnel management has a big difference, so maximizing the efficiency of strategic human resources management by identifying the differences between the two. The traditional personnel management conforms to the functional refinement and division of labor in the industrialized economic system, which allocates people as material resources and positions them in a costly way. The basic method of treating people is to start from the cost of resources, the use of efficiency and productivity level, avoiding investing more resource costs in the field of recruitment, selection, appraisal, personnel mobility, compensation, welfare benefits and personnel records. Therefore, it is often treated as a service work without professionalism and technology and becomes a service provider for other functional departments of the enterprise. Modernized strategic human resource management system is as the core of the whole staff's symbiosis, co-creativity, and harmonious coexistence with science and technology, humanistic spirit, collaboration,

consensus, and profit growth. Based on the material and spiritual needs of human resources, it takes the core principle of stimulating and driving people's subjective initiative and the degree of social cooperation, focuses on the development of human resources' asset and value-added attributes, and relies on systematic, strategic, long-term, and technical management in the fields of planning, forecasting, development, performance, and training, etc. Therefore, it is a series of resource allocation design, motivation and action activities centering on how human resources can improve the competitiveness of the enterprise in the market, and it is a project-type, global and systematic human resources operation system aiming at the strategic development of the enterprise [11].

3.4. Importance of strategic human resource management

Human resources are the main influencing factor for enterprise development and strategic plan development, and enterprises with abundant human resources are more competitive in the market. Various management activities such as resource development, business design and implementation of appropriate human resource management strategies can improve the competitive advantage of enterprises and manifest human rights. First, it enhances the execution of the enterprise. Production operations are the main activity of an enterprise to achieve its strategic objectives, which may be affected by employee competencies, corporate governance decisions, organizational culture, and other aspects. Therefore, when formulating and implementing strategies, enterprises will improve their corporate performance by developing strategic and systematic human resource management systems. Second, it enhances the core competitiveness of the firm. Strategic human capital is the source of creating the core competitive advantage of a firm, and the goal of human resource management is to enhance the core of the firm's competitiveness through acquisitions, and the maintenance of the firm's core competitiveness depends on the development and management of human resources. When the company's human resources become valuable, the company can enhance its human resource advantage through the development of strategic plan. Thirdly, it enables the company to obtain sustainable competitive advantage. The disadvantage of strategic human resource management is that it cannot gain competitive advantage in the short term, while the advantage is that it cannot be easily copied or imitated. Therefore, strategic human resource management emphasizes the concept of long-term management to create long-term competitive advantage for the company. Strategic human resource management focuses on implementing a range of human resource management activities in a systematic, strategic, and planned manner to encourage and support the achievement of goals and to continue to gain and maintain competitive advantage for the firm.

4. Challenges facing the landing of strategic human resources management in the new era

4.1. Inadequate human resource allocation

In order to control operating costs, some enterprises try to compress departments and positions as much as possible, and even require their employees to wear multiple jobs. Some employees in professional and technical positions spend a lot of energy on other work tasks and do not complete their own work to a high degree, the end result is that employees work inefficiently and perform poorly. In

addition, the human resource management of some enterprises lacks the basic concepts of labor economics and does not realize that the effective replacement of enterprise assets and human capital is an important method of controlling enterprise costs, which results in the wrong behavior of force resource allocation. For example, many enterprises in the development bottleneck, generally through layoffs to reduce corporate costs, this way to a certain extent can play a role in easing the enterprise out of difficulties, but in fact is not conducive to the long-term development of the enterprise. This kind of non-operational layoffs violates the principle of strategic human resource management, and the fundamental reason is that the enterprise's human resource planning is unreasonable.

4.2. Inadequate recruitment and training system

Recruitment is an important way for enterprises to obtain talents, and it is also an important content of strategic human resource management. However, in some enterprises, recruitment is random, usually when there is a job vacancy in the enterprise department or the demand for talents, only temporary recruitment, the lack of a complete recruitment plan, resume screening and interview process is too casual. The interview assessment process is missing, which ultimately leads to inefficient and ineffective recruitment, and it is difficult to recruit high-quality talents. In addition, the training system of some enterprises is too simple, unsystematic. The independence of each link is strong, but the connection is not close, and it pays little attention to the training needs of employees and the effectiveness of training. In the recruitment of new employees, the enterprise will generally carry out pre-employment training for employees. However, the content of the training is mainly the basic content such as the development history of the enterprise, rules and regulations, personal requirements instead of conducting specialized training on professional knowledge of the position, practical operation skills and related abilities. Training methods are also mostly lecture-based, which is the arrangement of training instructors on the staff of the top-down explanation. In this kind of employee training, it is difficult to attract the active participation of employees, did not allow employees to learn practical knowledge and work skills. It is difficult to effectively improve work efficiency and is not conducive to the achievement of the strategic objectives of the enterprise after training. Due to the lack of staff training system, the effect of corporate training is not ideal.

4.3. Inadequate Compensation and Benefit System

Reasonable compensation and benefits are both necessary to protect and enhance the life of employees, and also a reflection of the realization of their self-worth. However, in some enterprises, the lack of a sound compensation and benefits system has led to recruitment difficulties and high turnover rates. Compensation and welfare system is not sound in many aspects: firstly, the enterprise for the staff to issue benefits and the actual payment of employees do not match, the new employee probationary period is long, and the probationary period does not have "five insurance and one gold" and other welfare protection. Secondly, the salary model is unreasonable, such as "no responsibility base salary and no commission" salary model, and the salary adjustment is too arbitrary depending on the subjective judgment of the manager or human resources department. Thirdly, the salary level of the position is far below the average market standard. Due to the unsoundness of the enterprise's

remuneration and benefit system, it is not only unattractive to talents, but also difficult to retain general employees, and the remuneration below the market level is weakly attractive to social talents, which will adversely affect the efficiency and quality of the recruitment work.

4.4. Lack of systematic performance management

On the one hand, performance appraisal is treated as performance management. At present, due to the lack of in-depth understanding of performance management and performance appraisal, managers usually take the performance management appraisal system as a system of rewards and punishments for good or bad performance of employees and turn the performance appraisal system into the main management tool for punishing employees, rather than how to improve employee motivation through performance management. Performance management is a systematic process of work, through the setting of specific performance objectives and performance standards, and constantly guide employees to improve their work motivation. In addition, it is through the performance appraisal of the way to improve the ability of employees to enhance employee motivation to maximize the advantages of human resources and role. The ultimate goal of performance management is to improve the enthusiasm of employees and the operational efficiency of the enterprise. On the other hand, there is a lack of scientific performance index system. Performance evaluation is not based on the company's strategic perspective to design the indicator system, which cannot be analyzed between the indicators and the company's strategy to form a good relationship between the articulation. The lack of a rational system of performance appraisal indicators has led to structural contradictions within the enterprise and problems with the interface of work coordination between departments, as well as gaps between departments and posts The inconsistency of indicators prevents the competitive system from working in changing the strategic intent of each layer. At the same time, insufficient knowledge of performance management and lack of performance management practices prevented company managers from accurately distinguishing between performance indicators and criteria.

4.5. Lack of strategic human resource management awareness

Even though enterprises have set up human resources departments, their functions cannot be changed from personnel management to human resources management, the level of professional human resources management is not high, and human resources departments can only be responsible for simple personnel activities. Due to the lack of awareness of the importance of human resource management in the relevant departments, the role of strategic human resource management cannot be fully realized. The development of enterprises cannot be separated from people, and talent is an important guarantee for the survival and development of enterprises, so the scientific management of human resources is very important. But at present many enterprises in China in human resource management often exists in the end of the behavior, the management of the staff did not pay enough attention and will focus management on performance improvement rather than staffing and management. As a result, the performance improvement is not supported by the corresponding talent assurance, which ultimately leads to the performance improvement effect is not obvious. In addition, some managers of the enterprise simply think that personnel management is the management of people,

so only the personnel management is limited to the recruitment, ignoring the importance of the training, assessment and compensation and other aspects. This behavior seriously affects the quality of human resources management, and hinders the communication between human resources management and labor economics.

5. The Optimization Path of Strategic Human Resource Management of Enterprises Based on the Support of Labor Economics

5.1. Optimize human resources allocation

With the rapid development of social and economic development, the development of enterprises is also facing different environments, and human resource management needs to be based on the actual situation of enterprise development to make effective adjustments. In order to better carry out the scientific configuration of human resources management and effective adjustment of work, it is necessary to carry out a comprehensive optimization of the relevant personnel and organizational structure to ensure that it can better meet the distribution of human resources. When managing human resources based on the perspective of labor economics, it is only through the scientific configuration and optimization of the existing organizational structure and staff that we can effectively monitor the recruitment of talents and job allocation and promote the rapid development and utilization of human resources to the greatest extent possible. Enterprises can better stimulate the enthusiasm and enthusiasm of employees to work, give full play to their subjective initiative, create more substrate for the enterprise, and improve the enterprise's market competitiveness and social influence through scientific human resource management.

5.2. Formulate scientific recruitment plan and personnel training

Under the conditions of labor economics, strategic human resource management in the process of carrying out the work needs to be based on the actual needs of the company's positions to recruit the corresponding talent, the requirements must meet the recruitment conditions and the marginal cost of labor and marginal benefit of labor. After the arrival of the talent to the post in a timely manner for its pre-employment training, the enterprise based on the actual situation to develop training programs, which helps new employees to be able to have a clear understanding of the corporate culture and development goals in a relatively short period of time. Scientific training programs can help new employees to grow faster, which is the basic principle of labor economics in the investment of human capital, so as to provide reliable talent support for the healthy and long-term development of enterprises.

5.3. Optimize compensation and welfare system

Under the support of labor economics in the new era, enterprises must carry out scientific optimization and improvement of human resource management to achieve good development of enterprises, such as the improvement of the compensation system, the improvement of the incentive

assessment system and so on. In the formulation of the remuneration system, it must be a good performance mechanism and theoretical basis that the human resources department needs to be considered and paid attention to. Regardless of whether the level of remuneration is set at a high or low level, it needs to be established within the scope of the compensatory differences in labor economics, thus providing a more complete specification of the existing standards and requirements for performance pay in enterprises. In addition, the human resources department of the enterprise should also make use of the increased marginal cost, so that the marginal benefit generated by the production efficiency of the staff has a close relationship with the former, and finally to determine the efficiency of the enterprise.

5.4. Forecast labor market

In the process of strategic human resource management, not only should the enterprise internal human resources carry out scientific arrangements and management, but also need to give full consideration to the labor market situation. Enterprises make effective forecasts of the labor market for their own development, so as to optimize and improve their strategic human resource management. Because of the significant dynamics of the labor market, it is very difficult to predict the labor market. Especially when using the traditional forecasting methods, it often makes the forecast results not to meet the prediction goals and the poor accuracy, which directly leads to unsuitable strategies chosen in the process of human resource management, causing great economic losses to the organization. Therefore, the introduction and use of advanced technology is very important, such as cloud computing and big data technology [12], which can help enterprises better carry out the prediction of the labor market, improve the accuracy of human resources management, and also adjust to changes in the labor market.

5.5. Innovative human resource management concept

Under the scientific guidance of labor economics, the first thing to do in order to achieve effective implementation of strategic human resource management is to change the traditional management concepts, the existing labor resources for the scientific use, and enterprise development strategy and human resource management to make effective combination, so as to fully ensure that the enterprise can obtain higher economic benefits. And on this basis, the enterprise management and human resources leaders should also pay high attention to it, and on the basis of labor economics to realize the development of human resources management, in-depth analysis of the current problems and deficiencies in human resources management, so as to take effective measures to be dealt with scientifically, to promote the continuous improvement and updating of human resources management work, and reflect the main stronger New meaning.

5.6. Improve the cognitive ability of employees in economics

If an enterprise wants to be bigger and stronger in the fierce market competition, it must have a strong team that is willing to work for the development of the enterprise. In the talent team of the

enterprise, many new forces are almost all fresh graduates of college students, they not only have strong enough working energy, but also have a certain degree of self-knowledge, so in the actual work can put forward some excellent suggestions, the development of the enterprise has a constant flow of power, they can be said to be the enterprise's mid-life force. Therefore, enterprises should vigorously train this group, so that they can have a clear understanding of labor economics, in-depth understanding of the corporate culture and development process, and through training can also improve their ability and level of work. Therefore, enterprises can invite relevant professional understanding to carry out training on labor economics, which can not only help employees in their work to constantly try the theory-based methods of corporate revenue collection and expenditure reduction, but also provide better services for the enterprise and create more value. For the better employees in the enterprise, they can be formed into a small team to promote their internal exchanges, which can continuously improve their knowledge of labor economics, and also enrich themselves and improve their own ability.

6. Conclusion

Talent is the first resource for enterprises to promote high-quality development and the primary driving force for creating a high level of competitiveness. No matter the product advantage, technological advantage or market advantage, it must come from the talent advantage. As the owner of talents, enterprises must correctly manage and operate human resources, so that human resources can exert the maximum value in the process of enterprise development. The construction of a strategic human resource management system is to fully implement the idea of "people-oriented" and "talent leading development", upgrade and optimize the selection and employment mechanism, incentive mechanism, evaluation mechanism, attract, train, use and live the talents needed by the enterprise, and support the landing of the enterprise strategy. In short, human resource management is crucial in the process of enterprise development. But the current human resource management work is also facing great difficulties, especially after the emergence of labor economics, which has begun to become an important direction for the development of strategic human resources. This requires human resource management personnel to be able to focus on the labor market and the enterprise's own situation to carry out the management, and constantly improve the scientific nature of personnel recruitment, configuration and performance management, and thus improve the economic efficiency of enterprises.

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Responsibility of Intermediaries from the Case of Zeda and Amethyst

Yanchen Zhang [1*]

[1] East China University of Political Science and Law, Shanghai 200000, China
*Corresponding author Email: 1160219675@qq.com
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Abstract: The full implementation of the registration system provides more enterprises with listing opportunities, but also brings greater tests to the information disclosure system of the securities market. The change from the approval system to the registration system makes the intermediary assume the role of the gatekeeper of the information disclosure. Under the current attitude of the CSRC, issuers and intermediaries have been punished in an endless stream of news. Under false statements, how to punish the perpetrator and how to make up for the loss of the victim is the focus of the incident handling. The Amethyst Storage case and the Zeda Essence case, as recent typical cases, are also the first cases of fraudulent issuance and settlement of the science and technology board in China. The handling of the responsible personnel in the case and the application of the system of advance compensation and parties' commitment are of great value to the research on the liability of intermediaries.

Keywords: Intermediary organization; Misrepresentation; Liability boundary

1.Introduction

In 2023, the Issuing Department and Listing Department of the CSRC, together with the Shanghai Stock Exchange and the Shenzhen Stock Exchange, carried out the "Training Meeting for Intermediaries of Comprehensive Registration System", and the meeting mentioned that: strengthening the regulation of information disclosure, increasing the punishment, raising the cost of violation of laws and regulations, and complying with the information disclosure and gatekeeper responsibilities of the issuers. The severity of the penalties imposed by the SEC on the intermediaries in the Amethyst Storage case and the Zeda Essence case, as well as the severity of the accountability of the intermediaries, show the severe situation faced by the intermediaries in the process of public offerings under the full registration system.At the same time, the nature of the misrepresentation case also determines the nature of the case to deal with the settlement process of some special systems, such as the administrative party commitment system, the first pay system, the application of these systems to a large extent to protect the victims of fraudulent issuance, or to make the intermediary institutions from bearing heavier punishment, but also shows the intermediary institutions of the responsibility is very large. The case of Amethyst Storage and Zeda Essence as a sample, we should analyze the intermediaries and the responsibility of the parties, in order to find a better answer to the responsibility

of the intermediaries under the full registration system.

2. The Processing of the Two Cases and the Landing of Compensation

2.1 The Case of Amethyst Storage

2.1.1 Illegal Acts of the Parties and Punishment Decision

Firstly, Amethyst Storage failed to fulfill the relevant decision-making procedures and information disclosure obligations, and repeatedly violated the rules to provide large guarantees.Secondly, the disclosure of information related to capital restriction is inconsistent, and the information disclosure is untrue and inaccurate.

Based on legal provisions, the CSRC determined that the actual controllers and other organizations and instigations engaged in the above information disclosure violations constitute the illegal acts described in the first and second paragraphs of Article 197 of the Securities Law and make corrections, warnings, fines ranging from 500,000 yuan to 36.6852 million yuan. In addition, the company was also ordered to delist and entered the delisting consolidation period on June 8, 2023, and as of June 30, 2023, the company’ s shares have been traded for 15 trading days during the delisting consolidation period, and the delisting consolidation period has ended. The listing of the Company's shares was terminated and delisted by the Shanghai Stock Exchange on July 7, 2023.

2. 1.2 Application by Intermediaries for Application of the Party Commitment System

The sponsor agency China Securities Co., Ltd issued a verification opinion that Amethyst Storage does exist in violation of the guarantee situation. On September 3, 2021, the sponsor China Securities Co., Ltd made it clear in the special verification opinion of the “Shanghai Stock Exchange’ s Information Disclosure Regulatory Inquiry Letter for Amethyst Storage’ s semi-annual Report 2021” that at the end of June 2021, the company's monetary funds, in addition to the deposit used for issuing bank acceptance bills, were restricted funds. The rest of the funds do not exist due to pledge, guarantee and other circumstances resulting in limited funds. As the sponsor of Amethyst storage, the inconsistent answers of China Securities Co., Ltd are neither rigorous nor professional.

The CSRC has launched relevant investigations on China Securities Co., Ltd, Rongcheng Accounting Firm, Zhitong Accounting Firm, Guangdong Hengyi Law Firm and other intermediary institutions, and will deal with them according to law according to their diligence and responsibility in relevant practices, combined with their initiative to pay compensation in advance and apply for party commitment in securities and futures administrative law enforcement.

On May 27, 2023, China Securities Co., Ltd and Investment announced that the company, together with the accounting firm, Rongcheng Accounting Firm, and Guangdong Hengyi Law Firm formally established a special fund for the compensation of Amethyst event, with a total of 1 billion yuan. It is used to compensate investors for investment losses caused by the fraudulent issuance of amethyst storage and illegal information disclosure.

In addition, four intermediaries applied to the CSRC for the application of the Commitment system

for parties involved in securities and futures administrative enforcement, which was accepted by the CSRC in accordance with the law.

2.2 Handling of Zeda Essence Case

2.2.1 Illegal Acts of the Parties

Zeda Essence conceals important facts and fabricates material false content in the securities, and there are false records and major omissions in the disclosed 2020 Annual Report and 2021 Annual Report.

2. 2.2 Settled Through Mediation

On December 26, 2023, the SFC heard the investor v. Zeda and 12 defendants such as the actual controller, executives, intermediaries and other securities misrepresentation liability dispute was concluded by mediation. China Securities Small and Medium Investors Service Center,on behalf of 7,195 eligible investors received 280 million yuan in full compensation, covering 99.6% of the investors, of which a single investor received a maximum compensation of more than 5 million yuan, and the average person received a compensation of 38,900 yuan. Among them, Zeda Co., the actual controller Lin Ying, directly responsible for the supervisor should LAN bear the main liability for compensation. Other directly responsible persons of Zeda, intermediaries of securities issuance and their directly responsible persons shall be held responsible according to their respective degree of fault.

3.Application of Special Systems in Case Handling

3.1 Advance Compensation System

In recent years, China's capital market has carried out a series of practical explorations in the exercise of rights protection of small and medium-sized investors, and built up the First Compensation, Dispute Mediation, Ordinary Representative Litigation, Investor Protection Agency Representative Litigation and so on. Securities misrepresentation cases has a significant effect on investor compensation. According to the completed advance compensation cases, the indemnated investors account for more than 95% and the compensation amount accounts for more than 98%, and the claim time is greatly shortened and the claim efficiency is improved compared with litigation. Through the pre-payment system, investors can obtain compensation in a shorter period of time, avoiding lengthy litigation procedures.

Considering that the issuer's Amethyst storage and related personnel have limited compensation capacity, it may face difficult implementation through civil litigation. If investors obtain compensation through the way of advance compensation, it undoubtedly reduces the litigation costs of investors, increases the certainty and compresses the compensation time.

3.2 Commitment System for Parties Involved in Securities and Futures Administrative Enforcement

Commitment System for Parties Involved in Securities and Futures Administrative Enforcement, also

known as securities administrative settlement, refers to the securities regulatory agencies on suspected securities and futures violations of the unit or individual investigation, the party under investigation commitment to correct the suspected illegal behavior, compensation for the loss of the investors concerned, to eliminate the damage or adverse impact of the securities regulatory agencies recognized by the securities regulatory agencies, the party to fulfill the commitment of the securities regulatory agencies to terminate the case investigation of administrative law enforcement methods. The establishment of a party commitment system for administrative enforcement in the field of securities and futures, which introduces the contractual negotiation in the field of private law into the field of public power, is a major innovative step in the reform of the administrative enforcement system of the Securities and Futures Commission, reflecting the transformation of securities supervision from mandatory law enforcement to service-oriented law enforcement, and actively exploring a more diversified, flexible and rigid mode of governance.

3.2.1 Definition of the Amount of Commitment

According to Article 14 of the Measures for the Implementation of the Enforcement, the determination of the amount of the commitment fee shall comprehensively consider the amount of the party suspected of illegal acts that may be fined and the amount of illegal income confiscated. The loss suffered by the investor due to the suspected illegal acts of the parties and the law enforcement stage of the case at the time of signing the acceptance agreement.

In the case, the total commitment of the four intermediaries is 1,274,523,752 yuan, of which 1,085,585,416 yuan has been paid to investors through the early compensation procedure.

3.2.2 The Applicable Value of the Parties' Commitment System

Firstly,the interests of investors are protected to the maximum extent and with the highest efficiency.This case is dealt with by the combination of the commitment of the parties to the administrative law enforcement and the compensation in advance. According to the calculation of the insurance fund company, there were 17,471 damaged investors in this case, with a total loss of more than 1.097 billion yuan, and 16,986 investors received 1.086 billion yuan compensation in just two months, accounting for 97.22% of the total number of damaged people and 98.93% of the total amount of compensation.

Secondly, the parties suspected of violating the law were severely punished.In addition to the economic cost of about 1.275 billion yuan, the parties involved in the case also need to conduct self-inspection and rectification in accordance with the requirements of the China Securities Regulatory Commission, seriously investigate the responsibility of the responsible person and take internal disciplinary measures, strengthen the compliance risk control and management ability, and effectively improve the quality of practice. Among them, the relevant responsible personnel are not only subject to high fines, withholding and recovery of bonuses, dismissal, may not issue reports within a certain period of time and other internal disciplinary measures taken by their respective institutions, and the CSRC will take administrative regulatory measures according to the circumstances. The responsibility of the intermediary "gatekeeper" has been consolidated, and violations of laws and regulations have been

severely punished.

Thirdly, the effective combination of administrative law enforcement and civil compensation has been realized to improve the efficiency of law enforcement. Through the use of the comprehensive law enforcement method of the commitment of the parties to administrative law enforcement, on the one hand, the relevant law enforcement cases are quickly solved, the case is timely concluded, and the law enforcement efficiency is improved. On the other hand, the losses of investors are quickly compensated, the relevant civil disputes are timely resolved at the front end, the source of litigation is well managed, and the follow-up judicial resources are saved. It realizes the unification of administrative law enforcement and civil compensation, the unification of legal effect and social effect.

In addition, the application of the party commitment system in the amethyst storage case has restored market order in a timely manner and stabilized market expectations. Through disciplinary warnings for illegal actors and efficient compensation for investors, the market order is effectively maintained, the market environment is purified, and market expectations and confidence are stabilized.

3.3 The First Mediation of the Special Representative's Lawsuit Concludes the Case

The case of Zeda Essence is the second case of special representative litigation in China after the case of Kangmei Pharmaceutical case. The existence of the special representative litigation mechanism provides a strong support for the strengthening of post-event supervision under the framework of the registration system, effectively deterring illegal market behaviors, and escorting the comprehensive registration system. The purpose is to select “typical significant, bad social impact, with exemplary significance” cases, by the investor protection agency as the representative of the special representative litigation, while protecting the interests of investors, the subject of making false statements to severely punish, so as to deter potential offenders. For this reason, investor protection agencies should bring special representative lawsuits with a certain degree of continuity, in order to create psychological deterrence to potential offenders.

4. the deepening and boundary of the responsibility of intermediary institutions from the two cases

Intermediaries participate in the information disclosure process of securities issuance, and information disclosure is the front-end item and pre-work in the overall process of securities issuance. Under the registration system of false statements, intermediaries should “promote the front end from the back end, and promote the front end from the event.” In fact, in the context of the comprehensive registration system, regulators are constantly strengthening the supervision of intermediaries to improve the quality of business practice, and continue to consolidate the “gatekeeper” responsibility of intermediaries. In this regard, China Securities Co., Ltd in charge of investment banking business said in an interview, the company has done a deep introspection and serious rectification of the Amethyst storage project, to be paid through the first way to take the initiative to compensate for the losses of investors, to eliminate the related adverse impact, while continuing to improve the "three lines of defense", improve the whole process of quality control system, the establishment of a comprehensive risk management and compliance management system, and continue to improve the quality of practice, and effectively take up the capital market, "the gatekeeper" responsibility.

4.1 Evolution of the Role of Intermediaries under Different Issuance Review Systems

4.1.1 the Marginal Status of Intermediary Institutions under the Approval System

As the earliest issuance audit system in China, the approval system is a continuation of the high government color of the state-owned financial system in the capital market at that time, which was applied in the initial stage of China's capital market because of its adaptation to the environment at that time, and later it has been eliminated because of the market environment.

Under the approval system, the issuance of securities is completely arranged by the government, and the government arranges the quota, standards and conditions of securities issuance, and information disclosure is not a condition of securities issuance. Under the examination and approval system, the task of the intermediary is to help the issuer to make stock declaration. Because the key role of the intermediary is to check the information disclosure content, the role of the intermediary is very marginal under the examination and approval system.

4.1.2 The "IPO gatekeeper" under the Approval System Has Taken Shape

If the approval system is that the government handles the issuance of securities all by itself, then the government departments under the approval system have become parents who are willing to let go of their hands to a certain extent.

Under the approval system, the government department has the leading power of securities issuance, the substantive examination power and the final approval power, and at the same time, it reduces the workload of the government department under the approval system by strengthening the information disclosure. The government will take the initiative to make a substantive review and judgment on the issuer, which is based on the information disclosed by the issuer under the supervision of the intermediary agency. The compliance of information disclosure and the content of disclosure become the audit object of the audit authority. Due to the supervisory function of intermediaries in the process of information disclosure, the quality and efficiency of securities issuance under the approval system are better than those under the approval system. Under the approval system, the role positioning of intermediary agencies as "IPO gatekeeper" has begun to take shape, but under the approval system, the administrative leadership of government departments is still in an absolute position, and the intermediary agencies only assume the role of auxiliary.

4.1.3 the Depth of the Establishment of "IPO gatekeeper" under the Registration System

The driving force of the reform of the registration system is to return the choice to the market, strengthen the market constraints and the rule of law constraints. Instead of having the government, the parent, make the choices ahead of time for the market. The core of the registration system is information disclosure. Compared with the information disclosure system under the approval system, the information disclosure under the registration system is more central and dominant. It can be said that the legal regulation under the registration system is a series of institutional arrangements with information disclosure as the core. The core status of information disclosure also means that the

participation of intermediaries has been greatly enhanced. Its responsibility is an important part of the responsibility system under the registration system.

Under the registration system, the issuance of securities has become a formal review, and the government organs no longer conduct strict substantive review on the quality of securities, but the object of formal review is the issuer’ s information issued by intermediaries.As the main body of securities issuance, the issuer bears the primary responsibility for all kinds of problems that occur in the issuance process. The intermediary, as an auxiliary to the issuer and an intermediary organization with regulatory functions, plays a role in the process of securities information disclosure. Under the registration system, the intermediary has completed the profound establishment of its role, which is a well-deserved “IPO gatekeeper”.

4.2 The Responsibility of An Intermediary Institution Is Reasonably Assumed

In essence, the intervention of intermediaries is to "increase credit" for issuers to issue credit, and through their professional activities, they help issuers reduce the investigation costs and transaction costs of investors on issuers' principal credit. The mechanism of post-hoc accountability for intermediary institutions should also be determined by the actual function and positioning of intermediary institutions in the market. The return of intermediary institutions is an important basis for the smooth implementation of the registration system, and is also an important entry point and focus for the high-quality development of the capital market.

4.2.1 Clear Responsibility Content and Details

In this context, it is particularly important to vigorously promote the high-quality development of intermediary institutions. The high-quality development of securities management institutions must achieve clear strategy, accurate positioning, strong professional strength, effective management and control, sustainable development and excellent culture. Strengthening the construction of the rule of law and tightening the responsibility of intermediary institutions is a major focus of registration system reform. At present, the legal constraints on intermediary institutions have been basically complete. In practice, there are still many details to be clarified in scientific allocation of intermediary responsibility.

From the history of global securities trading, the initial securities trading actually has no intermediary underwriting, sponsorship, audit or lawyer verification and other professional services, which brings the problem that the credit cost of transactions between market entities is too high, affecting the formation of the best game transaction price. In essence, the intervention of intermediaries is to “increase credit” for issuers to issue credit, and through their professional activities, they help issuers reduce the investigation costs and transaction costs of investors on issuers' principal credit. The mechanism of post-hoc accountability for intermediary institutions should also be determined by the actual function and positioning of intermediary institutions in the market. The key to clarify the responsibility of intermediary institutions is to distinguish the “special duty of care” and “general duty of care” that intermediary institutions should bear. In terms of how to determine accurately, the judicial interpretation stipulates that each intermediary agency should bear the special duty of care in its own professional field, and the general duty of care is not in the professional field. However, in practice,

which is the duty of special care, which is the duty of general care, very test the wisdom of judicial practice.At present, a very important issue is to clarify accounting responsibility and audit responsibility. In practice, the most common intermediaries - securities firms, law firms and clubs - may be the most bitter, and a prominent problem is that accounting responsibility and audit responsibility have yet to be clarified. The quality of intermediary professional services is related to the quality of listed companies and securities, and then to the interests of investors. Too light responsibility will encourage the violation of law, too heavy responsibility will make the intermediary institutions overwhelmed, only clear how to identify the responsibility of intermediary institutions can balance the protection of investors and the long-term development of intermediary institutions.

4.2.2 Clarify "expert obligations" and "non-expert obligations"

As for the fault determination of intermediaries, Article 11 of the Securities Act of 1933 stipulates that non-professionals shall rely on the truth of relevant statements after "reasonable investigation" for the content without professional opinion support and the professional opinions issued by professionals themselves; A lay person does not need to make a "reasonable investigation" into a professional opinion or official document issued by a professional, only to prove that there are no reasonable grounds to believe and do not believe that there is a misrepresentation or omission; A professional shall not be liable for any misrepresentation other than that of his professional opinion. China's Judicial Interpretation of False Statements, recent regulatory rules and judicial precedents all show a tendency to distinguish expert liability from non-expert liability. However, in practice, how to identify the identity of underwriters and sponsors, and how to judge the reasonable trust standard of intermediary institutions for professional opinions of other intermediaries remains to be explored.

4.2.3 Clarify the Boundaries of Responsibilities Between Intermediaries

When determining the responsibility of intermediaries, we should also pay attention to the differences in their duties and obligations under different identities. Taking securities companies as a typical representative, according to relevant regulations, the continuous supervision obligations of securities companies as sponsors and financial advisers are obviously different from their responsibilities in the issuance and restructuring stages. In the stage of issuance and reorganization, sponsors and financial advisers have the obligation to check and verify the information of listed companies, while in the stage of continuous supervision, sponsors and financial advisers only have the obligation to review the general information disclosure of listed companies in time, but do not have the obligation to prudently check or guarantee the truth, accuracy and completeness. In the bond misrepresentation dispute, there have been cases in which the bond trustee is listed as a co-defendant and requires the liability for misrepresentation. The bond trustee is neither a bond underwriter nor a bond service institution, and its focus is to perform its duties fairly and safeguard the rights and interests of the holders, and it has no obligation to prudently check the information disclosure documents of the issuer during the duration of the bond. The trustee management report usually makes clear statements and hints that the quoted content is not verified. It is expected that judicial cases will clearly define the identity and responsibilities of the bond trustee and provide useful guidance for the orderly

development of the bond market.

5. Conclusion

The change from the approval system to the registration system makes the intermediary assume the role of the gatekeeper. Under the current attitude of the CSRC, the intermediary institutions should actively implement self-management, ensure that they perform their duties in the process of securities listing audit.And they should actively settle the claim that loss of the investor group. Give full play to the advantages of the system of advance compensation and commitment of the parties to solve the problem. The implementation of advance compensation by intermediaries, has reduced the litigation costs of investors, increased the certainty, compressed the compensation time, and largely eliminated the adverse effects of the malignant events of false statements. But on the other hand, the responsibility boundary between the four intermediaries is still not very clear. If it needs to be further clarified, it still needs to be further discussed from the performance content of the intermediary, fault or not, and expert obligations, which is also an important part of the identification of the responsibility of the intermediary agencies in other cases.

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Rescarch on Perfecting the Forced Delisting System of List Companies

WAN Ya-jun [1*]

[1] School of Law, Anhui University of Finance & Economics, Anhui Bengbu 233041, China
*Corresponding author Email: 2112160649@qq.com
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Abstract: The forced delisting system, functioning as a mechanism for the "export" of listed companies, is pivotal in fostering market dynamism and optimizing resource allocation within the securities market byenforcing the principle of survival of the fittest. which is directly related to whether China's securities market can realize the orderly and timely clearing pattern. However, there are still difficult and slow delisting problems in China's securities market. The fundamental obstacle is that the standard of forced delisting is generally loose, the forced delisting procedure is lengthy and the connection is not smooth. Drawingon the experience of developed capital markets outside the region, it is suggested that in terms of entity system, the number of market makers should be increased to fully reflect the will of investors, and forced delisting standards for trading should be strictly formulated to strengthen market standardization, while the weight of financial profit standards in delisting standards should be reduced, and the scope of application of non-quantitative standards should be expanded to comprehensively combat delisting evasion. In terms of procedures and systems, shorten or directly cancel the consolidation period to speed up the delisting speed; At the same time, improve the transition rules after forced delisting to ensure the smooth transition of the company to be delisted.

Keywords: Listed company; Forced delisting; Market maker; Non-quantifiable; Standard; Finishing period; Cohesion rule

1.Introduction

In 2018, the introduction of the Science and Technology Board's stock issuance registration system necessitated enhancements to the delisting framework. In response to facilitating the registration system's implementation and fostering the normalization of delisting, the Securities Law underwent amendments in March 2020, significantly streamlining the mandatory delisting process. This included the abolition of listing suspensions and resumptions, delegating the authority to regulate mandatory delisting scenarios to the stock exchange, and introducing a dedicated section to underscore the safeguarding of investors' rights and interests. In 2022, the Stock Exchange revised the Stock Listing Rules for the twelfth time, further intensifying the disclosure of delisting risks and refining the delisting metrics. Subsequently, in 2024, the China Securities Regulatory Commission issued the Opinions on the Strict Implementation of the Delisting System, calling for further standardization of mandatory delisting criteria. While the reform of the registration system has resulted in increasingly refined mandatory

delisting arrangements, practical challenges persist, such as evasion loopholes, inadequate post-delisting support mechanisms, and difficulties in safeguarding the legitimate interests of affected small and medium-sized investors. These challenges pose impediments to the effective implementation of a normalized delisting mechanism. In recognition of the numerous practical issues within China's mandatory delisting system, this paper aims to conduct a thorough analysis of the current delisting framework and propose practical solutions to further refine the mandatory delisting system and safeguard the rights and interests of small and medium-sized investors.

2. Current situation of forced delisting of listed companies

2.1 Current situation of forced delisting system of listed companies

The relevant provisions of the delisting system in China were first seen in the Company Law in 1993 and the Securities Law in 1998. After several amendments, the content of the forced delisting system in the Company Law was transferred to the Securities Law. At present, China has formed a listed company delisting mechanism mainly composed of the relevant provisions of the 2020 "Securities Law", several opinions issued by the CSRC, and the listing rules of the stock exchange.

2.1.1 The provisions of the Securities Law on forced delisting

According to the relevant provisions of the current Securities Law, if a listed company violates laws or regulations or violates the delisting situation stipulated by the stock exchange, the stock exchange makes a delisting decision according to the corresponding norms and standards, and the company to be delisted has objections, it has the right to apply to the review institution of the stock exchange for review. It can be seen that the formulation of specific rules on the Forced delisting of listed companies is authorized to the stock exchange, and the company to be delisted can only be relieved through the internal channels of the exchange if it refuses to accept the Forced delisting decision.

2.1.2 SFC rules on forced delisting

In 2024, the CSRC issued the Opinions on the Strict Implementation of the Delisting System (hereinafter referred to as the Opinions), which once again emphasized that the delisting system is the key basic system of the capital market. Under the pressure of the recent market fraud wave and the downward fluctuation of the stock market, the Opinions require the exchange to scientifically set strict diversified forced delisting standards in combination with the listing conditions of enterprises, and gradually broaden the diversified exit channels to accurately achieve "should be retreated". And accelerate the revision of stock listing and related business rules, release and implement as soon as possible, and increase the clearance of "zombie shells" and "black sheep"[1].

2.1.3 Stock exchange rules on forced delisting

Based on the comprehensive reform of the registration system, the stock exchange, authorized by the upper Law and guided by the principles of the CSRC, formulated specific forced delisting rules according to the actual situation of the market and regulatory requirements. On April 30, 2024, based on the "Opinions" issued by the CSRC, the Shanghai and Shenzhen Stock Exchanges effectively revised and completed the relevant delisting rules and issued the "2024 Delisting New Regulations". Under the guidance of the Opinions, the new delisting regulations in 2024 further strictly regulate the standards for Forced delisting, which are mainly reflected in the following three aspects: expanding the scope of

application of Forced delisting for major violations; Tighten the financial delisting standards, improve the operating income delisting indicators of loss-making companies, and increase the delisting efforts of poor companies; Three new normative delisting situations have been added.

3. Problems of forced delisting system of listed companies

3.1 The forced delisting standard of listed companies cannot play the functions of prevention and punishment

3.1.1 Lack of market maker number standards

Market maker refers to the institution or individual authorized by the stock exchange to trade stocks, bonds, etc., in the securities market, which usually has strong financial strength and market reputation. In 2022, China introduced the system in the science and technology board, but it was not introduced in the A-share market. Some people believe that this is because the core role of the market maker system is to increase market trading volume, the turnover rate of stocks in the A-share market is already high, and there is no need to introduce it, and by comparing the turnover rate and liquidity of the science and Technology board market and the A-share market, it is found that these two indicators of the science and Technology board significantly exceed the average level of the A-share market, which indicates that the introduction of the system is not only to improve liquidity. From the analysis of market positioning, the science and technology board is in A sense the experimental field of A-share market reform, and its introduction of the market maker system has more far-reaching significance and consideration.

Although the liquidity of the A-share market is very high, the scale of China's A-share market continues to grow, and has reached more than 5,000, from the perspective of the entire A-share market, the two-level differentiation is obvious, some stocks are marginalized, and local liquidity is insufficient. In view of this, it is necessary for China to introduce this system in the A-share market, and it also has good conditions for introduction. Considering that the number of market makers can well reflect the liquidity characteristics of listed companies, and liquidity to a large extent can reflect the degree of investor favor, and investor favor is undoubtedly the basis of listed companies. Therefore, the introduction of this system is of great significance for improving the operating efficiency of the A-share market and the competitiveness of listed companies. Nasdaq's continuous listing standards stipulate the number of market makers. In the capital market sector, no matter what standard is used to obtain listing qualification, the number of market makers is required to be no less than 2; in the global Select sector and the global market sector, the number of market makers is required to obtain listing qualification according to the shareholders' equity standard, and the number of market makers is required to be at least 3. To qualify for listing on the market capitalization basis or on the total assets/gross revenue basis, a minimum of 4 market makers are required. With the in-depth implementation of the comprehensive registration system of China's securities market, the degree of marketization has been further strengthened, in this context, in order to better improve the delisting system, improve market efficiency and protect investors' rights and interests, it is necessary to consider increasing the number of market makers in the forced delisting standards.

3.1.2 The standard of forced delisting of trading class is lax

The standard of forced delisting of transactions is mainly divided into four aspects, which are the

closing price, the total market value, the turnover and the number of shareholders. The exchange also stipulates specific standards according to whether the listed company issues A shares, B shares, and both A shares and B shares. The following takes the forced delisting standard of A-share trading as an example to analyze. Total market value provisions: "The total closing market value of stocks for 20 consecutive trading days is less than 500 million yuan", this value setting does not match the status quo of China's securities market, according to Wind statistics show that as of January 1, 2024, the number of A-share listed companies is 5,346, with A total market value of 84 trillion yuan, and an average market value of 15.712 billion yuan. Listed companies with a market capitalization of more than 1 billion accounted for 99%. Considering the actual situation of the current A-share market, when stipulating the market value delisting standard, only requiring the total market value to be less than 500 million yuan is too low, which lacks realistic pressure on listed companies and cannot play the early warning role that should be had. Practice has also proved that since the standard was released in 2020, no company has been forced to delist for falling below this standard.

In terms of the number of shareholders: "The number of shareholders of the company is less than 2000 for 20 consecutive trading days", this standard is also too loose, first of all, China's A-share market small investor base is large, with the development and growth of the size of the securities market, the number of investors will continue to increase. At present, few companies are forced to delist because of the standard, and there will be fewer in the future, so it is necessary to combine the actual situation of China's securities market, and moderately raise the delisting standard of the number of shareholders. In addition, the period of 20 consecutive trading days can also be criticized. When the number of shareholders of the listed company may fall to the forced delisting standard, within 20 trading days, the listed company can take various means to manipulate the number of shareholders. As long as it does not meet the requirements of "continuous", it can easily resolve the delisting risk, at the same time, in terms of supervision, the stock exchange and the CSRC are difficult to monitor the change in the number of shareholders in real time, and there are obstacles to the investigation and collection of evidence of illegal regulation and control behaviors, which bring costs to supervision, and cause the dilemma that inferior listed companies should not withdraw. In view of this, it is necessary to shorten the period of 20 trading days in view of the actual situation of the stock market.

3.1.3 Financial profitability criteria are overweighted

Compared with mature markets such as Nasdaq, the delisting standards of Shanghai and Shenzhen exchanges focus more on financial criteria such as net profit and operating income. According to Wind data, in 2023, among A-share delisting companies, financial delisting accounted for 44%, and trading delisting accounted for 40%, at the same time, there were four new cases of major illegal delisting in 2023, the specific reasons are also related to false revenue and profit. Therefore, to some extent, stock exchanges tend to formulate and adopt financial standards in the forced delisting standards, which may be due to the frequent occurrence of financial fraud in practice. At the same time, it may also be due to the fact that stock exchanges pay more attention to the consideration of financial standards in the delisting standards, leading to the delisting of more companies. It is undeniable that financial standards are the most intuitive indicators reflecting the operating conditions of a company. For stock exchanges, the delisting of listed companies based on quantitative financial standards is undoubtedly the most

cost-effective. Compared with the normative information disclosure review and the illegality identification of major violations, it saves a lot of human and material resources, and also reduces the conflicts caused by differences of opinion.

However, everything has a dual character, and quantitative financial standards are also the "cost performance" pursued by listed companies to avoid delisting. For example, the current delisting rules tighten the financial standards and improve the operating income standard of loss-making companies, increasing it from "100 million yuan" to "300 million yuan", and adopt the combination standard of "total profit/net profit/withholding non-net profit + operating income". Compared with the revision before, the standard coverage is wider, the surface appears more demanding, but in fact, listed companies are forced to delist to touch the double standards of profit and operating income, as long as the operating income is not less than 300 million yuan, even if not profitable, do not have to worry about delisting. In this regard, the stock exchange has limited the scope of the main business income of listed companies to a certain extent, in order to combat the inflated operating income of listed companies through non-main income channels. However, in practice, listed companies often avoid delisting by rapidly increasing their operating income by confirming the recovery of accounts receivable in advance. In recent years, the phenomenon of listed companies deliberately whitewashing financial statements through contractual arrangements and manipulating income by adjusting completion schedule is not uncommon, and it is difficult to completely eradicate in the future. In addition, delisting standards tend to focus on financial standards can also lead to "mistake", evaluation of a company's good or bad, can not only based on whether the loss, need to consider its sustainable business ability and potential. Moreover, the delisting rules adopted by China have led to the delisting of some companies that have touched major illegal delisting standards according to other standards. Among other standards, financial standards are the easiest to identify, which to A certain extent explains why financial delisting accounts for the largest proportion in the A-share market.

3.1.4 Non-quantified standards are not comprehensive

Non-quantitative standards, that is, qualitative standards, can not be measured by specific values or indicators, mainly involving corporate governance, illegal and other aspects. In the stock market of most countries and regions, the delisting criteria of listed companies usually include quantitative criteria, such as continuous losses, low stock price, insufficient number of shareholders, etc. These criteria are clear and clear, making it easier for regulators and investors to assess delisting risks. However, non-quantitative standards tend to be more subjective and flexible.

China's current delisting system in terms of qualitative standard requirements, mainly includes in the standard category, major illegal category two aspects. In 2020, the Shanghai and Shenzhen Stock Exchanges issued the "Shanghai and Shenzhen Stock Exchange Listing Rules (December 2020 Revision)", which added qualitative standards for "information disclosure" and "reporting false records". In 2023, in order to meet the requirements of the implementation of the comprehensive registration system, the listing rules of Shanghai and Shenzhen exchanges were revised again, and the qualitative standards in terms of normative categories mainly include false records, changes in share capital/equity, forced dissolution, bankruptcy reorganization, etc. The category of major violations stipulates the qualitative criteria for damages to market order, national interests and public interests, and is refined through a

"qualitative + quantitative" approach. For example, listed companies involved in major illegal acts stipulated in Article 9.5.1 (a), When the amount falsely recorded in its operating income, total profit or net profit reaches the maximum proportion of the amount in any year, it will trigger a major illegal forced delisting. In 2024, the stock exchange amended the stock listing rules under the guidance of the Delisting Opinions to expand the scope of major illegal delisting and increase the situation of continuous fraudulent delisting in terms of qualitative standards. Compared with 2020, China's non-quantitative standards have been more comprehensive, but compared with foreign mature markets, the coverage of qualitative standards is narrow. A-share corporate governance and sustainable operation standards need to be further improved, such as the development of more comprehensive and strict non-quantitative standards for forced delisting, involving shareholder voting rights, related party review and supervision, compensation distribution, audit committee management, independent directors and other aspects.

3.2 The forced delisting procedure rules of listed companies cannot achieve efficient delisting of companies to be delisted

3.2.1 The forced delisting period in the procedural rules is long

In April 2024, the exchange's revision of the stock listing rules did not involve the Forced delisting procedure rules for listed companies. At present, the basic process of Forced delisting for listed companies is as follows: triggering delisting conditions - delisting risk warning (except for trading) - termination notice and decision letter - Delisting consolidation period (except trading) - delisting - board transfer trading or bankruptcy reorganization. The delisting arrangement period refers to the period of time for liquidation and arrangement of a company after receiving the notice and decision of the stock exchange to terminate its listing.

According to the delisting process, the listed companies entering the delisting consolidation period have gone through the "delisting risk warning" and other links, which last for 2 to 6 months, and the exchange has served the notice of termination of the listing and the decision, on this basis to set a 15-day delisting consolidation period is cumbersome. The delisting period is intended to provide an opportunity for the company to rectify its business and financial position and avoid being delisted. For this purpose, it may lead to two situations: first, in this very short period of time, the company is difficult to achieve the boom, even if the success of profit, it is also a small profit, if based on this small profit listing again, it can not rule out the company in the short term to face the possibility of delisting again, in the long run, is not conducive to market stability; Second, it is easy to lead companies to take risks to commit fraud, causing greater disruption and resistance to the market and supervision. There are also views that the delisting consolidation period is set up to provide the final exit channel for shareholding investors, but also to give investors willing to hold the last buying opportunity. The author believes that this view is too idealistic, because once the company is issued by the exchange to terminate the listing notice and decision, it is necessary to disclose information in a timely manner, for such a buyer's market at this time, the investors who buy in the delisting period are completely out of the mentality of speculation and gambling. In practice, there is often a "light" before the official delisting of the company, such as diamond retreat, the stock price soared on the fourth day of the delisting consolidation period, and the cumulative increase in 11 trading days is as high as 91.3%. This surge is obviously unreasonable,

and the delisted stocks have lost the substantial investment value, which is completely the speculative gambling behavior of investors relying on capital advantages. Seriously disrupting market order.

3.2.2 The connecting rules of the turntable are not smooth

The stock market should be an "in and out" market. Compared with foreign mature markets, the delisting rate of listed companies in China is low. The reason lies in the lack of perfect rules for connecting the board after delisting. At present, the exchange only stipulates that delisted companies need to transfer their shares to securities trading venues such as the New Third Board for transfer, but the specific operating rules are not clear.

In practice, listed companies often transfer to the new third Board market after delisting from the main board market. This one-size-fits-all approach makes it more convenient for regulators to supervise them in the later stage, but it is very unfavorable for companies in need of financing and small and medium-sized investors whose rights and interests are damaged. Because the New Third Board market has been in the "mud" for a long time, the liquidity is very poor, which undoubtedly increases the difficulty of these companies in financing, but also makes the protection of the rights and interests of small and medium-sized investors face greater challenges, so the new third Board actively introduces the market maker system. However, due to the small scale of market makers, high threshold, narrow scope of targets and other reasons at this stage, the function of the market maker system has been restricted, and the liquidity of the new third Board market has not been greatly improved. The author believes that the root cause of the poor liquidity of the new third Board lies in the small investor base of the new third Board and the lack of investor confidence in this sector. According to the relevant regulations, the threshold for individual investors to buy and sell shares in the delisting consolidation period is: more than 2 years of trading experience and no less than 500,000 yuan of securities assets per day (within 20 trading days before the opening of the authority). The restrictions were originally intended to protect investors and reduce the risk of trading in delisted sectors. However, with the deepening of the marketization of China's securities market, investors should be given greater transaction autonomy and lower threshold restrictions. At the same time, strengthen the information disclosure of delisted companies to enhance investor confidence. In addition, the current transfer rules may lead to huge transfer costs, according to the provisions of the exchange, after the company to be delisted, the exchange has agreed to revoke the probability of the decision to terminate the listing of the company's shares, but at this time the company's shares have been transferred to the new third Board and other securities trading venues, need to go through the share confirmation, registration and other procedures, resulting in a waste of resources.

4. Experience in designing legal system of forced delisting of foreign listed companies

4.1 Diversified Forced delisting standards ensure the quality of stock listed companies

4.1.1 NASDAQ: a combination of quantitative and qualitative delisting standards

In view of the diversified characteristics of the market, NASDAQ combines quantitative analysis and qualitative evaluation methods, and formulates differentiated delisting standards according to the uniqueness of each market. First, trading standards, including minimum number of investors, minimum public shareholding, number of registered and active market makers; The second is the going concern

standard, including pre-tax income (pre-tax profit from continuing operations)[2], total assets, total cash flow, etc.; The third is compliance standards (mainly from the perspective of the company's internal governance), including independent directors, solicitation agents, audit committees, executive salaries, voting rights, financial disclosure and other requirements. Among them, transaction standards and going concern standards are quantitative standards, and compliance standards are qualitative standards[3]. It can be seen that although NASDAQ adopts a combination of quantitative and qualitative methods to formulate delisting standards, it still prefers qualitative standards in general and pays more attention to the weight of enterprise compliance. At the same time, the delisting standards of the three market segments within NASDAQ also have clear differentiation arrangements, taking the shareholder equity standard as an example, the global select market needs to reach at least $55 million, the global market needs to reach at least $30 million, and the capital market only needs to reach $5 million[4].

4.1.2 LSE: The exchange takes subjective consideration based on the actual situation

The biggest difference between the LSE and other global trading markets is that when dealing with the delisting of listed companies, it often makes subjective considerations based on the actual situation. This subjective approach reflects LSE's emphasis on market flexibility and situational awareness, making its delisting system more in line with the actual needs of the market.

The UK rarely sets quantitative standards for forced delisting, and LSE has certain discretion when considering whether to force a company to delist from the exchange, and will comprehensively consider a variety of factors, including but not limited to the company's financial condition, business performance, governance structure and future development prospects. As the LSE Main Board forced delisting situation provides: The Exchange reserves the discretion and flexibility to adjust the standards in certain areas where appropriate, breaches of this standard shall be at the sole discretion of the Exchange. London's Alternative Investment Market (AIM) also has rules: securities trading is chaotic; Listed companies violate rules; It is necessary to protect investors; In order to maintain the reputation of market integrity, in the above cases, the exchange will conduct a comprehensive assessment of the overall situation of the company and make a subjective judgment based on the market environment and the interests of investors, and may suspend the trading of securities when it considers appropriate. According to the Financial Services and Markets Act (FSMA) issued by the British government, the suspension and Forced delisting criteria of the London Stock Exchange mainly include financial status, fee payment, illegal and other aspects. When the listed company does not meet the relevant continuity obligations or does not comply with the regulations, the exchange may order it to suspend trading or force it to delist[5].

4.2 The efficient forced delisting procedure promotes the normalization of delisting of listed companies

4.2.1 NASDAQ : Hearing system program design

Forced delisting is initiated by the exchange, and when the listed company triggers the delisting conditions, NASDAQ will take measures such as independent review, suspension rectification and delisting. The specific procedure is as follows: First, NASDAQ issues a delisting warning. On a daily basis, NASDAQ's Eligibility Department monitors listed companies in real time and compiles a list of companies that violate the continuing Listing Rule. Once it is determined that a listed company has violated one or more of the Continuing Listing Rules, NASDAQ will send a delisting warning to the listed company. If the

listed company violates the compliance or operating standard version of the Continuing Listing Standard, it shall submit a compliance improvement plan to Nasdaq for review and regain compliance with the NASDAQ Continuing Listing Standard within 45 days after receiving the Nasdaq delisting alert. The grace period generally does not exceed 180 days (up to 360 days can be extended). If the listed company does not meet the liquidity indicators in the listing standards for 30 consecutive days, after receiving the Nasdaq delisting warning, the listed company shall make corrections within 180 days and meet the liquidity standards for at least 10 consecutive days. If the continuous listing requirements are not met, the listed company can automatically drop the board or enter the delisting process. If a listed company disagrees with a decision such as a delisting warning, it can submit a written application and appeal the decision step by step, the basic process is: Listing Eligibility Committee - Hearing panel - Listing and Suspension Review Committee - the US Securities and Exchange Commission.

4.2.2 TSE: Steping program design

TSE is the central stock exchange market in Japan. It is divided into two markets, the first board and the second board, according to the scale and liquidity standards. Newly listed companies are listed on the second board market. Will be relegated to the second board market.

The TSE has established a step-by-step procedure for forced delisting of listed companies. The basic process is: When the exchange finds that the company does not meet the conditions for continued listing, it first carries out special treatment (similar to China's ST system) and gives a certain grace period to test whether it can meet the listing standards again. If it meets the standards again, it revokes the previous treatment; if it does not meet the standards, the TSE will make further restrictions on it. They are classified as "supervisory shares" (which are handled by the Supervisory Office within the exchange). If the company still fails to meet the standard within the time limit, the stock exchange will designate it as a "consolidation stock" and confirm that the listed stock has been forcibly delisted. After designating it as a "consolidation stock" (which is processed by the consolidation office of the stock exchange), the company will be delisted after a three-month consolidation trading period. Most of the listed companies forced delisting is due to corporate restructuring and liquidity reasons, by the executive director of the stock exchange to make the company forced delisting decision, the decision can be effective once made, is not actible[6].

5. Improving the legal system of forced delisting of listed companies

5.1 Strengthen the preventive and punitive functions of the forced delisting standards for listed companies

5.1.1 Add the Standard of market makers to the forced delisting criteria

According to the relevant provisions of China's Securities Law, the way of listing securities can adopt other legal ways approved by the securities regulatory agency. Market maker trading is one such way. Therefore, the introduction of market maker system in the A-share market, and then increase the number of market makers in the continuous listing conditions of listed companies, there is A definite upper law basis.

On January 19, 2024, the Beijing Stock Exchange released the results of the 2023 annual evaluation of market makers, showing that by the end of 2023, the Beijing Stock Exchange had a total of 16 market

makers and filed 198 single market making stocks, involving 88 stocks. Compared with the market makers before joining, the average daily turnover rate of the underlying stocks increased by 34%, the relative bid-ask spread and intraday volatility decreased by 18% and 7%, respectively. It can be seen that relative to the number of listed companies, the team of market makers in China still needs to be expanded. Therefore, it can guide more qualified market making institutions to participate in the construction of the New Third Board market, and increase the number of market makers in the continuous listing conditions. This move can also provide experience for increasing the number of market makers in the forced delisting standard of the A-share market. The market makers are highly specialized investment institutions, while the majority of investors in China are individuals, and the investment blindness is large. Adding the number of market makers to the forced delisting criteria can provide a good model for individual investors. However, some scholars believe that the implementation of the market maker system in A-shares may induce market manipulation and adversely affect the market order. At the same time, market makers need to hold 1% to 5% of the shares of the listed companies responsible for market making, the required capital scale is huge, and the volume of securities companies in China is limited, so the implementation of this system faces challenges. After more than 30 years of development, China's securities market has grown into an influential international market, and the supervision mechanism of the securities market has been continuously improved. Large securities firms such as Sinolink Securities and CITIC Securities have already met the market making requirements[7]. Nevertheless, in view of the limited number of market makers at present, it is recommended to moderately relax the number of market makers, while maintaining a prudent attitude to the qualification requirements of market makers.

5.1.2 Strictly formulate forced delisting standards for transactions

The new rules for delisting in 2024 raise the expected market value, income and other standards of the third set of listing standards. The estimated market value of the third set of standards will be increased from "$8 billion" to "$10 billion", and the operating income in the latest year will be increased from "$800 million" to "$1 billion". It will also increase the market value standard of the main board A shares (including A+B shares) from the current "300 million yuan" to "500 million yuan". The author believes that, on the one hand, the improvement of the standard shows that the exchange and the CSRC have realized that the current standard setting cannot keep up with the development of the securities market, but on the other hand, compared with the increase of the expected market value of the listing standard, the increase of the forced delisting market value standard is too low, which does not meet the coordination principle of the delisting of the securities market. From "300 million yuan" to "500 million yuan", in practice, it can not put forward substantive requirements for the governance level of listed companies, for the realization of "should retreat", and promote the formation of an orderly and timely liquidation pattern. It is suggested to refer to the European Union's Digital Market Law on the determination of whether an enterprise has a significant impact on the internal market, based on a certain proportion of the overall average market value of listed companies each year, the last place is eliminated, and the listed companies that do not meet the market value and do not meet the conditions during the grace period are forced to delist. Regarding the forced delisting standard for the number of shareholders, we can learn from the relevant experience of Nasdaq, and set different requirements for

the number of shareholders according to different standards for companies to obtain listing qualifications. At the same time, we can also further refine the requirements for the number of shareholders according to different industries of listed companies. For example, for hot investment fields such as new energy industry and artificial intelligence, differentiated requirements for the number of shareholders can be set up according to different industries of listed companies. The standard for the number of shareholders required for Forced delisting should be raised accordingly. Taking into account the differences between China's securities market and the US market, investors in the US securities market are mainly institutions, while small and medium-sized investment accounts for a large proportion in China, so accordingly, in the number of shareholders, a certain forced delisting standard setting, China's current standard should be further improved, according to the disclosure of shareholder data of 1117 listed companies since 2024. The number of shareholders is in the hundreds of thousands, so it is proposed to raise the standard of forced delisting shareholders to more than 3000.

5.1.3 Downplay the role of financial profitability criteria in delisting criteria

Downplaying the status of financial profit standards in delisting standards does not mean that financial standards are not important in delisting supervision, but aims to better play the synergistic effect of various forced delisting standards and avoid financial standards becoming the only dominant one. This is not an abatement or weakening of the existing rules, but a further clarification and strengthening of the applicable principles of the forced delisting standard for listed companies.

Regarding the application of forced delisting standards, China's current system follows the principle of "first touch first apply", but the author believes that this principle has drawbacks in specific practice, and should adopt the principle of core touch for delisting arrangements. The so-called core touch means that the delisting is not simply decided according to the order of the listed company's forced delisting standards. Instead, a comprehensive consideration should be given to the various standards touched by the listed company, and the delisting should be arranged according to the standards that best reflect the core requirements of the stock exchange's supervision of listed companies. Taking Shanghai Zhongchang Big Data Co., LTD. (hereinafter referred to as "Zhongchang") as an example, according to the full text of the delisting Zhongchang (600242) announcement, in 2023, Zhongchang was punished by the CSRC for information disclosure is not timely and incomplete, and in 2022, the company's performance loss was 599 million yuan, and shareholders illegally reduced their holdings of stock companies. Financial statements were issued unable to express opinions, in 2021, the company's performance loss of 472 million yuan, the company's director and the original person in charge of the subsidiary during the term of office suspected of embezzlement of the company's interests, embezzlement of funds were criminally filed. It can be seen that the shareholders and management of the company have been violating laws and regulations for a long time. In the end, the stock exchange terminates the listing based on the principle of "first touch, first apply" and the audit report that its accounting report is issued and cannot express its opinion. However, the application of this standard to delisting does not have a strong warning effect on listed companies. The author believes that the core reason for the delisting of the company is a major illegal behavior, involving the violation of major information disclosure and seriously affecting the listing status, and it should be delisted according to the forced delisting provisions, so as to better highlight the regulatory function of the forced delisting

system, and also better warn potential illegal companies.

5.1.4 Broaden the coverage of non-quantified standards

In terms of non-quantitative Forced delisting standards, foreign exchanges have more extensive discretion. In addition to paying attention to information disclosure, bankruptcy liquidation, and fee payment, non-quantitative standards such as shareholder voting rights, executive compensation, audit system, and related transactions are also included in Forced delisting standards, and there are some non-quantitative standards. For example, the compliance of the corporate governance structure, the soundness of the internal control mechanism, and the performance of the company's social responsibility are directly related to the interests and rights protection of investors[8].

In contrast, China's current non-quantitative Forced delisting standards are limited to the reporting authenticity, Forced dissolution, bankruptcy reorganization and other companies' relatively shallow operating conditions, and do not involve the shareholder voting rights within listed companies, the compliance of governance institutions, financial audit and whether there are connected transactions. In practice, in order to avoid the fate of forced delisting, many poorly operated listed companies transfer their non-performing assets to other companies through related party transactions, and then inject high-quality assets to beautify their financial reports[9]. In addition, some listed companies change their control rights through the acquisition of related parties or buyback of their own shares, as well as through the private placing price lower than the share price or book value, which will seriously damage the legitimate rights and interests of investors. Therefore, it is necessary to address the above-mentioned speculation to avoid delisting. In the subsequent revision of the listing rules, the coverage of non-quantitative standards will be further expanded to better combat the behavior of listed companies to avoid delisting.

5.2 Multiple measures to promote the forced delisting procedures of listed companies to achieve efficient delisting

5.2.1 Shorten the consolidation period before forced delisting

In view of the difficulty of delisting in China's securities market and the low delisting rate, we can consider learning from the practice of mature markets[10], and suggest limiting the consolidation period before Forced delisting to within 7 trading days or further cancel the setting of the consolidation period for touching all Forced delisting standards. According to the above discussion, no matter whether the setting of the consolidation period is to save the listed company or to protect the investors who hold shares, the setting of the consolidation period cannot play a good expected effect. In the former case, a warning period of as much as two to six months will not allow listed companies to regain the conditions for listing, and a short period of consolidation after the decision to delist will not have much effect. On the latter, after the delisting decision of the exchange was disclosed, new investors entered the market completely out of a speculative mentality, they rely on capital advantages, low absorption and high selling to earn the difference, and finally cashing out, which is bound to bring more investors greater losses.

From the perspective of liability, the shareholding investors of the delisted company should seek self-equity relief according to the fault principle. If the stock is delisted due to market factors, investors should bear the profit and loss risk. If the delisting is due to human factors such as managers and major

shareholders, the responsible party should make compensation, instead of selling shares through the secondary market within the consolidation period. It is also doubtful whether it can inherit the former investor's right to claim damages from the responsible party. Even if there is a right of inheritance in theory, the maneuverability in practice is low, which undoubtedly constitutes unfair treatment for new investors. On the contrary, the existing investors have received full compensation from the responsible party, and the setting of the delisting consolidation period has lost the necessity.

5.2.2 Improve the rules for connecting the board after forced delisting

With the implementation of China's registration system and the implementation of the normalized delisting mechanism, some companies with good operating conditions may also be forced to delisting due to the "1 yuan delisting" standard, and there are also enterprises that choose to actively delisting. The delisting of these companies does not mean their demise, and they may be re-listed in the future. Therefore, it is important to provide perfect transition places for these enterprises, ensure their smooth transition and maintain market order. In order to achieve this goal, we can strengthen the construction of the new third board market[11], reduce the entry threshold, increase the investor base, and improve market liquidity. At the same time, increase the information disclosure of the delisted companies on the New third Board to ensure that investors understand the operating conditions and reasons for delisting. For companies delisted for different reasons, set differentiated information disclosure requirements, especially for major illegal enterprises, the highest standards should be set to reveal risks[12]. The implementation of these measures not only provides a good transition place for delisted companies, maintains market order, but also enhances the attractiveness and competitiveness of the New Third Board market, and provides investors with diversified investment options.

At the same time, we can learn from NASDAQ's practice of transferring to the board after delisting, and NASDAQ listed companies have three choices after forced delisting: transferring to other exchanges, over-the-counter trading system or privatization. Different exchanges have different listing standards, and delisted companies can transfer to the board if they meet the requirements of other exchanges. After China's listed companies are delisted from the main board, they may still meet the listing standards of other trading plates, such as the science and Technology version and the GEM board, which have relatively low requirements. Therefore, if the company to be delisted meets the requirements of these sectors, it can directly apply for transfer to the board without re-listing. The two sectors are highly liquid, helping investors limit their losses. It is suggested that in the later stage, when improving the transfer rules of the company to be delisted, we should break through the single delisting path and formulate the transfer rules from the main board to the GEM or the science and technology innovation board. In addition, in order to avoid wasting resources, the buffer period should be reserved for the consideration of the Listing Committee to avoid the reversal of procedures when optimizing the connection rules of the transfer board.

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Application of Adaptive Machine Learning Systems in Heterogeneous Data Environments

Xubo Wu [1*], Ying Wu [2], Xintao Li [3], Zhi Ye [4], Xingxin Gu [5], Zhizhong Wu [6], Yuanfang Yang [7]

[1] Independent Researcher, USA
[2] University Maine Presque Isle, USA
[3] University of Miami, USA
[4] Elevance Health USA
[5] Northeastern University, USA
[6] Google LLC, USA
[7] Southern Methodist University, USA
*Corresponding author Email: franky.wu2333@gmail.com
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Abstract: This paper explores the application and effectiveness of adaptive machine learning systems in heterogeneous data environments. With the diversification of data sources and types, traditional machine learning systems face numerous challenges, especially in data processing and model adaptability. Adaptive machine learning technologies optimize the capability to handle multi-source heterogeneous data by dynamically adjusting learning algorithms and model parameters, enhancing model accuracy and robustness. Research through theoretical analysis and multiple experiments demonstrates the effectiveness of adaptive systems in various application fields such as healthcare and finance, highlighting their advantages in complex data scenarios such as high noise and missing data. Future research will focus on improving model interpretability, optimizing large-scale data processing capabilities, expanding cross-domain applications, and strengthening data security and privacy protection to promote the widespread application and development of adaptive machine learning technology.

Keywords: Adaptive Machine Learning Systems; Heterogeneous Data Environments; Data Quality; Data Integration; Deep Learning; Generalization Ability

1. Introduction

1.1 Research Background and Significance

With the rapid development of digital technology, the explosive growth of data volumes and diversification of sources bring new opportunities and challenges to the application of machine learning technology. The ubiquity of heterogeneous data environments, especially the vast differences in data formats, structures, and sources, complicates the design and optimization of machine learning systems.

The core challenge lies in effectively processing and utilizing heterogeneous data to improve the adaptability and robustness of machine learning models. Adaptive machine learning systems, as an efficient tool to address these issues, hold significant theoretical and practical value. In the context of rapid information development, data from various industries such as text, images, audio, and video are increasing, creating a diversified data ecosystem. Each data type contains unique information and value. Integrating and harnessing these data's potential is key to enhancing machine learning system performance. While traditional machine learning methods excel in handling specific types of data, their performance is often limited in diverse data environments. Adaptive machine learning technology, by dynamically adjusting model parameters and structures, not only optimizes the processing effects of a single data source but also provides stable performance across data types, significantly enhancing model generalization capabilities. Data quality and integrity also directly affect the output quality of machine learning. In heterogeneous data environments, inconsistencies in data collection often result in missing data, excessive noise, and outliers. If unaddressed, these issues can lead to poor model training outcomes or even misguided decisions. Adaptive machine learning systems enhance data quality and ensure efficient and accurate model training and application by incorporating advanced data preprocessing techniques, such as anomaly detection and data imputation. Moreover, data integration technology is key to effectively utilizing heterogeneous data. Data from different sources and formats must be fused through reasonable integration strategies to fully utilize information. Adaptive machine learning systems facilitate this process by designing flexible data integration architectures and algorithms, further expanding machine learning applications and depth.

1.2 Analysis of Current Research

Application Fields	Research Outcomes
Healthcare	Utilizing deep learning technologies like CNNs and RNNs for disease prediction and diagnosis has improved accuracy and speed.
Finance	By using semi-supervised learning and transfer learning techniques, the accuracy and adaptability of financial fraud detection have been enhanced, reducing false positives.

Table 1: Application of Adaptive Machine Learning Systems in Heterogeneous Data Environments

Existing research shows that adaptive machine learning technology has achieved significant results in various fields, especially in scenarios with high data diversity and quality challenges. In critical areas such as healthcare and finance, the application of technologies like deep learning and transfer learning not only enhances data processing precision but also strengthens the model's adaptability to new situations. Moreover, by addressing issues in real-world applications, such as inaccurate data labeling, label noise, and imbalanced distributions, researchers have proposed solutions like semi-supervised learning and multimodal fusion, providing strong support and broad prospects for the development of adaptive machine learning systems.

2. Fundamentals of Adaptive Machine Learning Systems

2.1 Introduction to Machine Learning Concepts

In recent years, with the advent of the big data era, machine learning technology has been widely applied across various domains. In heterogeneous data environments, traditional machine learning systems often face challenges such as uneven data distribution and inaccurate feature extraction. Therefore, adaptive machine learning systems have become important tools for addressing these issues. Adaptive machine learning systems are characterized by high flexibility and strong adaptability, allowing automatic adjustments based on different data environments. In heterogeneous data environments, adaptive machine learning systems can automatically select appropriate learning algorithms and models based on the characteristics and distribution of the data, thus enhancing system performance and accuracy.

Compared to traditional machine learning systems, adaptive machine learning systems have higher generalization capabilities and adaptability. Through continuous learning and adjustment, adaptive machine learning systems can quickly adapt to new data environments, enhancing model stability and generalization ability. In heterogeneous data environments, the types and characteristics of data may change, but adaptive machine learning systems can automatically recognize and adapt to these changes, ensuring model accuracy and efficiency.

Overall, adaptive machine learning systems have significant application prospects in heterogeneous data environments. By enhancing system flexibility and adaptability, adaptive machine learning systems can better cope with complex data environments, achieving more precise predictions and more effective decisions. In the future, as big data continues to grow, adaptive machine learning systems will play an increasingly important role, driving the application and development of machine learning technology across different fields.

Features	Traditional Machine Learning Systems	Adaptive Machine Learning Systems
Flexibility	Lower, difficult to adapt to new data environments	High, can automatically adjust based on the data environment
Adaptability	Limited, slow to respond to data changes	Strong, can quickly recognize and adapt to data changes
Learning Algorithm and Model Selection	Fixed, usually requires manual selection of appropriate algorithms and models	Dynamic, the system automatically selects the most suitable algorithms and models
Performance and Accuracy	May be limited by uneven data distribution and inaccurate feature extraction	Through continuous self-optimization and adjustment, performance and accuracy are enhanced

Generalization Ability	Generally poor, especially when the data environment changes	Higher, through continuous learning and adjustment, model stability and generalization ability are improved
Continuous Learning and Adaptation to New Environments	Weaker, needs to be retrained or adjusted to cope with new environments	Strong, the model can continuously update to adapt to new data environments

Table 2: Comparison of features between traditional machine learning systems and adaptive machine learning systems

2.2 Overview and Workflow of Adaptive Machine Learning Systems

Adaptive machine learning systems are a collection of highly flexible intelligent algorithms designed to cope with changing data environments. These systems adjust learning strategies and model parameters in real-time to adapt to data characteristics and task requirements, especially effective in handling heterogeneous data environments, showing excellent performance and generalization ability.

The workflow of adaptive machine learning systems includes several key steps to adapt to continuously changing data environments:

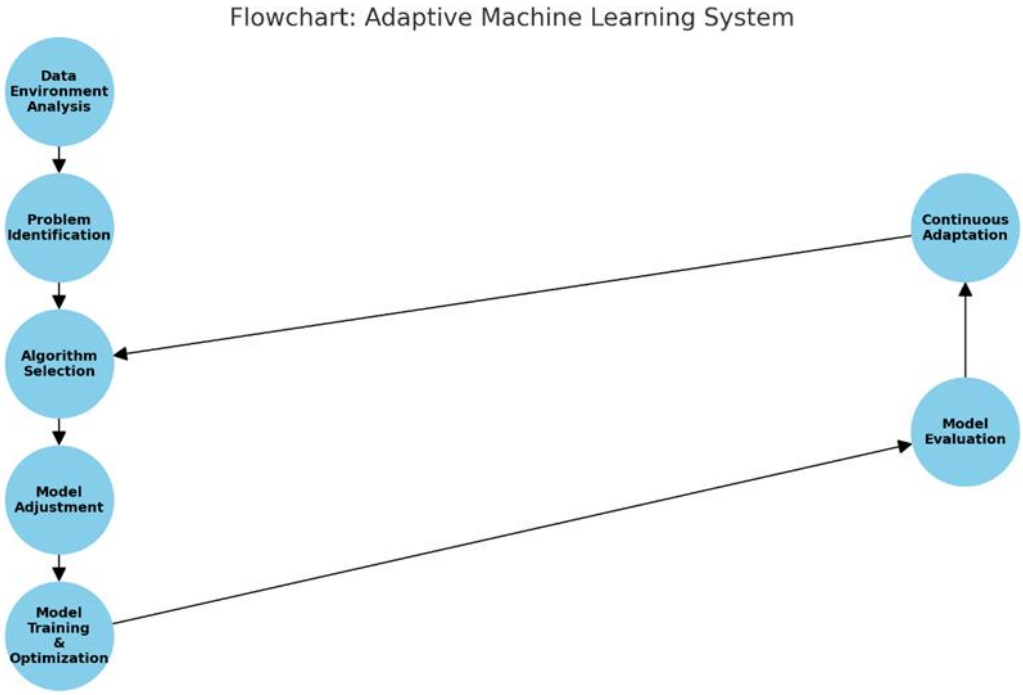
- **Data Input:** The system first receives heterogeneous data from various sources. This step is crucial as the diversity and timeliness of data are the basis for the effective operation of adaptive systems.
- **Dynamic Feature Selection:** When processing input data, the system dynamically selects the most relevant features based on the current data characteristics. This process uses machine learning algorithms to identify which features are most effective in predicting future events or outcomes, thereby enhancing model prediction accuracy.
- **Real-time Algorithm Optimization:** Based on the selected features and continuously received new data, the system adjusts its algorithm parameters in real-time. This step ensures that the model can quickly adapt to changes in data distribution, optimizing the decision-making process.
- **Continuous Model Learning:** The system continuously learns from new data, updating the model through constant training. This continuous learning mechanism allows the model to maintain high adaptability and robustness when facing unknown data or changes in the environment.
- **Model Output:** After processing through the above steps, the model outputs its prediction results. The output not only reflects learning from historical data but also considers the impact of real-time data, thus more accurately reflecting the current environmental state.

Through this series of steps, adaptive machine learning systems maintain efficiency and accuracy in variable data environments, adapting to various operational demands. These systems are widely applied in fields such as personalized medicine in healthcare, real-time trading systems in financial markets, and threat detection in cybersecurity.

The core advantage of these systems lies in their exceptional adaptive capability, automatically adjusting their algorithms based on immediate data changes. This includes adjusting algorithm parameters, optimizing learning strategies, and dynamically adjusting model structures. This flexibility significantly enhances model predictive accuracy and robustness on unknown data.

Adaptive machine learning systems also face several challenges in practical applications, including high computational demands, the complexity of real-time data processing, and maintaining model stability in dynamic environments. Additionally, the maintenance and updating of systems require complex technical support to ensure continuous performance optimization and adaptability.

In the commercial and research sectors, adaptive machine learning systems have been applied in several important areas, such as personalized treatment plans in healthcare, real-time trading systems in financial markets, and threat detection in cybersecurity. These applications demonstrate the tremendous potential and value of adaptive systems in solving practical problems.



2.3 Analysis of Heterogeneous Data Environments

2.3.1 Mathematical Models and Theoretical Analysis

Applying adaptive machine learning systems in heterogeneous data environments requires a deep understanding of the automatic adjustment process of model parameters from both mathematical models and theoretical perspectives. This section will detail how adaptive machine learning systems identify and adapt to heterogeneity in data and how this adaptability is reflected in the model learning process.

In adaptive machine learning systems, the model parameter update rule is typically represented in the form of gradient descent:

$$\theta_{new} = \theta_{old} - \alpha \nabla J\left(\theta\right)$$

where θ_{new} represents the updated model parameters, θ_{old} represents the previous model parameters, α is the learning rate, and $\nabla J(\theta)$ is the gradient of the loss function $J(\theta)$ with respect to θ. The following provides a detailed derivation of this update rule's mathematical foundation and explains

its applicability in heterogeneous data environments.

2.3.2 Basic Principles of Gradient Descent

Gradient descent is an iterative optimization algorithm aimed at minimizing the loss function to find the optimal values of model parameters. The basic idea is to update the parameters in the direction of the negative gradient of the loss function, as the gradient indicates the direction of the fastest increase in function values, while the negative gradient indicates the fastest decrease.

The loss function $J(\theta)$ is typically expressed as:

$$J(\theta) = \frac{1}{m}\sum_{i=1}^{m} L\left(\hbar_{\theta}(x^{(i)}), y^{(i)}\right)$$

where $\hbar_{\theta}(x^{(i)})$ is the model's prediction, $y^{(i)}$ is the true value, L is the loss function, and mmm is the number of samples. To minimize $J(\theta)$, its gradient needs to be calculated:

$$\nabla J\left(\theta\right) = \frac{\partial J\left(\theta\right)}{\partial \theta}$$

Assuming the loss function is mean squared error (MSE), i.e.:

$$J\left(\theta\right) = \frac{1}{m}\sum_{i=1}^{m}\left(\hbar_{\theta}(x^{(i)}) - y^{(i)}\right)^2$$

Taking the gradient with respect to θ:

$$\nabla J\left(\theta\right) = \frac{\partial}{\partial \theta}\left(\frac{1}{m}\sum_{i=1}^{m}\left(\hbar_{\theta}(x^{(i)}) - y^{(i)}\right)^2\right)$$

Using the chain rule, we obtain:

$$\nabla J\left(\theta\right) = \frac{2}{m}\sum_{i=1}^{m}\left(\hbar_{\theta}(x^{(i)}) - y^{(i)}\right)\cdot\frac{\partial \hbar_{\theta}(x^{(i)})}{\partial \theta}$$

Assuming $\hbar_{\theta}(x) = \theta^T x$, then:

$$\frac{\partial \hbar_{\theta}(x)}{\partial \theta} = x$$

Thus, the gradient is expressed as:

$$\frac{\partial J\left(\theta\right)}{\partial \theta} = \frac{2}{m}\sum_{i=1}^{m}\left(\hbar_{\theta}(x^{(i)}) - y^{(i)}\right)x^{(i)}$$

2.3.3 Derivation of Update Rule

Using gradient descent to update parameters:

$$\theta_{new} = \theta_{old} - \alpha\nabla J\left(\theta\right)$$

Substituting the gradient, we get:

$$\theta_{new} = \theta_{old} - \alpha\left(\frac{2}{m}\sum_{i=1}^{m}\left(\hbar_{\theta}(x^{(i)}) - y^{(i)}\right)x^{(i)}\right)$$

2.3.4 Selection of Learning Rate α

The learning rate α is a key parameter that controls the pace of parameter updates. Choosing an appropriate α is crucial for the convergence and stability of the model. In heterogeneous data environments, the characteristics of different data sources may cause significant differences in the size

and direction of gradients. Therefore, adaptive models often use dynamic adjustment methods for the learning rate, such as:

- **Adaptive Learning Rate Algorithms:** Algorithms like AdaGrad, RMSprop, and Adam adjust the learning rate for each iteration based on the history of gradients.
- **Learning Rate Scheduling:** Gradually reduces the learning rate during the training process to enhance the model's convergence performance and stability.

2.3.5 Adapting to Heterogeneous Data Environments

In heterogeneous data environments, the features and structures of data vary greatly, potentially coming from different data sources or domains. Traditional fixed-parameter models struggle to handle this complexity, while adaptive machine learning systems can dynamically adjust model parameters according to the characteristics and structures of the data, better adapting to different types of data. This adaptability is reflected in the following aspects:

- **Dynamic Adjustment of Learning Rate:** Adjusts the learning rate based on data characteristics and gradient changes, enhancing the model's convergence speed and stability in different data environments.
- **Parameter Updates:** Uses the gradient descent algorithm to update model parameters in each iteration based on data distribution and features, better adapting to the current data environment.

Through the above formula derivation and theoretical analysis, it is evident that adaptive machine learning systems have significant advantages in handling heterogeneous data environments, effectively coping with changes in data characteristics and structures, and improving model generalization capabilities and predictive performance.

2.3.6 Mathematical Visualization

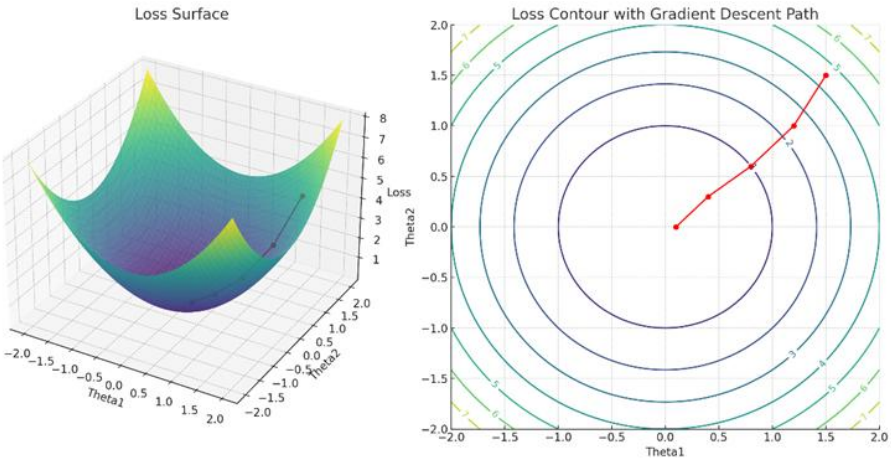
To more intuitively demonstrate the learning effects of adaptive machine learning systems in handling heterogeneous data and the model parameter update process, we use mathematical visualizations for analysis.

Loss surface graphs can display the shape of the loss function $J(\theta)$ in parameter space, as well as the path of parameter movement during the gradient descent process. This allows us to visually observe the model optimization process and the adaptive model's ability to find optimal solutions in the loss function space.

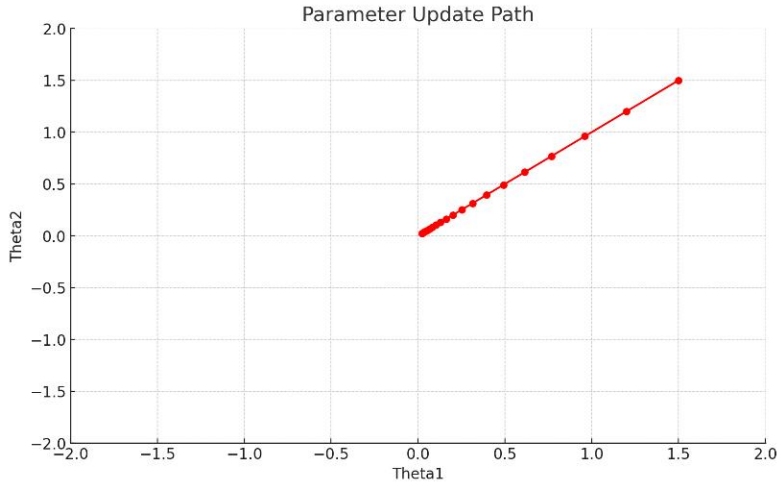
Assuming our loss function is in quadratic form:

$$J(\theta) = \theta_1^2 + \theta_2^2$$

We can plot this function's surface graph in parameter space and show the path of gradient descent.



Parameter update dynamics graphs display how the parameters θ_{new} $and\theta_{\text{old}}$change during the model training process. Through these visualizations, we can observe how parameters are gradually optimized in each iteration to adapt to the characteristics of heterogeneous data.



2.3.7 Theoretical Optimization Strategies

To make adaptive machine learning systems more effective in handling heterogeneous data, several optimization strategies can be employed, such as dynamically adjusting the learning rate, using different parameter initialization methods, and adopting advanced gradient optimization techniques (e.g., Adam or RMSprop). These strategies help the model maintain a stable learning process when facing data source diversity.

By understanding and designing adaptive machine learning systems through mathematical models and theoretical analysis, we can better comprehend and design these systems, enabling them to perform more effectively when dealing with data of different characteristics and structures.

2.4 Experimental Design and Analysis

To further verify the effectiveness of adaptive machine learning systems in heterogeneous data environments, we designed a series of experiments using synthetic datasets and open-source datasets. These experiments aim to demonstrate system performance under different heterogeneous conditions and analyze key factors in the learning process.

2.4.1 Experimental Design

Data Set Construction:

To simulate heterogeneous data environments, we constructed three types of datasets, each with different feature distributions, noise levels, and missing data ratios. Specific types include:

- **Uniform Distribution Data:** This dataset is designed to simulate an ideal statistical learning environment where feature values are uniformly distributed with a low noise ratio (signal-to-noise ratio of 20 dB), to assess the model's optimal performance under standard conditions.
- **Gaussian Mixture Data:** Reflects real-world conditions, with data features generated according to several different Gaussian distributions, each representing a data subpopulation with varying means and variances. This design tests the model's ability to handle data internal diversity.
- **Noisy and Missing Data:** In this dataset, feature values are randomly introduced with a high proportion of noise (signal-to-noise ratio of 10 dB) and 30% random missing values, to examine the model's robustness in handling low-quality data.

Model Configuration:

We compared adaptive machine learning models with traditional fixed-parameter models. Adaptive models dynamically adjust learning rates and weight decay parameters according to data characteristics, while fixed-parameter models use uniformly set hyperparameters in all experiments.

Statistical Methods and Performance Metrics:

- **Performance Evaluation:** Models are assessed using accuracy, F1 scores, and AUC values. Additionally, loss function values for each model on different datasets are calculated to evaluate their learning efficiency and stability.
- **Statistical Tests:** Repeated measures ANOVA is used to analyze the impact of different datasets on model performance and the significant differences between adaptive models and fixed-parameter models.

2.4.2 Experimental Results and Analysis

Experiment 1: Uniform Distribution Data

The adaptive model quickly converges, with loss decreasing from 0.9 to 0.2, indicating effective optimization. Statistical tests show that there is a statistically significant difference in performance between the adaptive model and the fixed-parameter model on uniform distribution data ($p < 0.05$).

Experiment 2: Gaussian Mixture Data

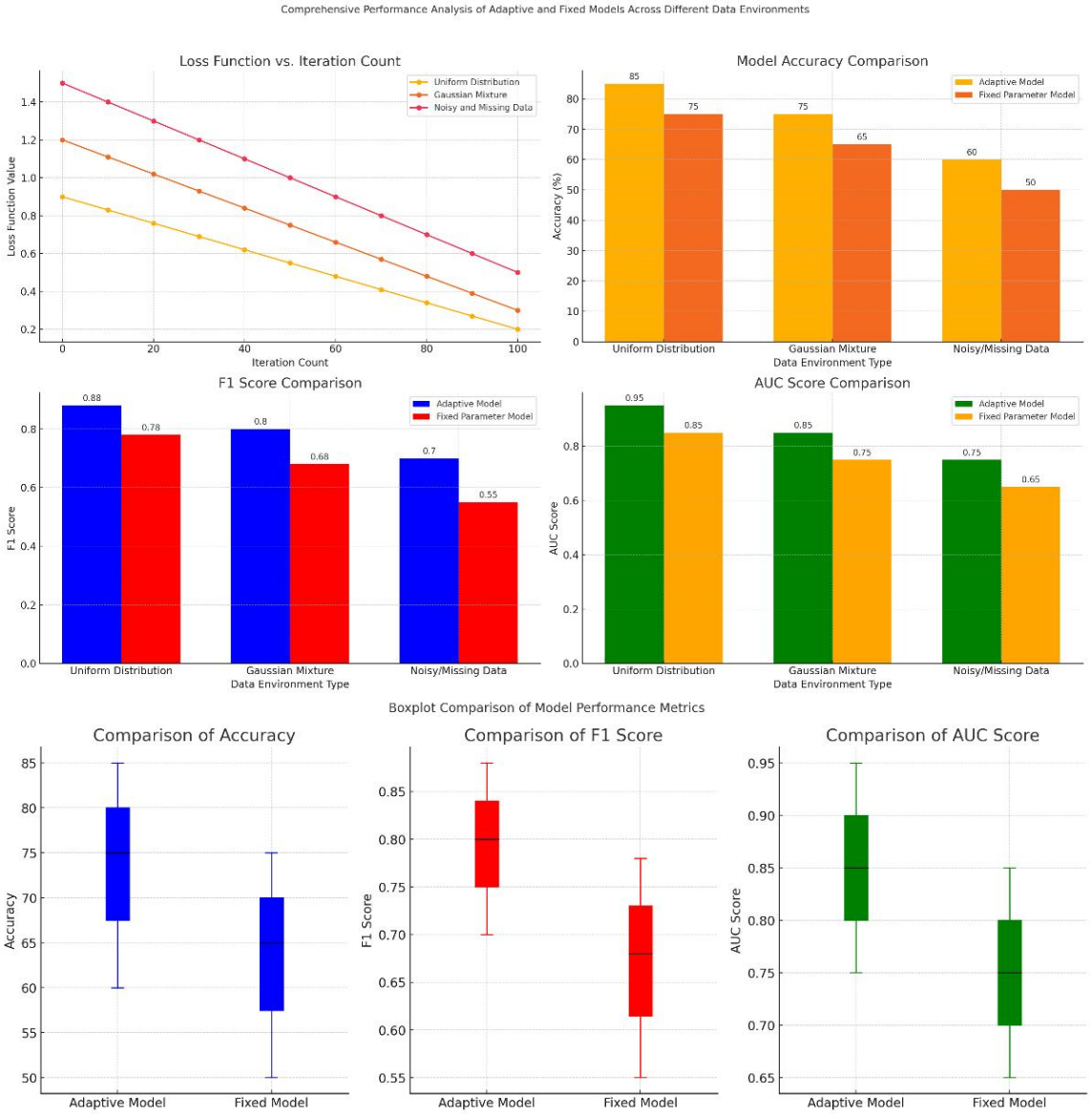
The adaptive model demonstrates better performance in multimodal distributions, with loss slowly decreasing from 1.2 to 0.3. Performance is superior to that of the fixed-parameter model, especially in terms of higher classification accuracy and F1 scores among data subpopulations.

Experiment 3: Noisy and Missing Data

In dealing with noisy and missing datasets, the adaptive model exhibits significant robustness, with loss decreasing from 1.5 to 0.5. Repeated measures ANOVA analysis indicates that the performance of the adaptive model is significantly better than that of the fixed-parameter model in this low-quality data environment ($p < 0.01$).

Through these simulated experiments and data analysis, we can conclude that adaptive machine learning systems demonstrate higher performance and better adaptability when handling heterogeneous data. These experimental results not only support the effectiveness of adaptive machine

learning technology but also provide theoretical and practical bases for future applications in more complex data environments.



3. Conclusion

This study, through in-depth analysis and experimental verification, demonstrates the efficiency and flexibility of adaptive machine learning systems in handling heterogeneous data environments. Adaptive machine learning technology can adjust learning strategies and model parameters in real-time in variable data environments, optimizing model performance and enhancing accuracy and stability. Additionally, our experimental results emphasize the potential applications of adaptive systems in critical areas such as healthcare and finance, as well as their robustness in complex scenarios involving noisy and missing data.

Future work can explore and expand in several areas:

- **Model Interpretability:** Further research into the decision-making processes of adaptive machine learning models and the development of more transparent model explanation tools to enhance model acceptability and trust.

- **Large-scale Data Processing:** Optimize adaptive machine learning algorithms for big data environments, improving system processing capabilities and efficiency, ensuring high performance in scenarios with vast amounts of data.
- **Cross-domain Applications:** Explore the application of adaptive machine learning technology in more fields, such as environmental science, intelligent manufacturing, and the Internet of Things, expanding its application scope and impact.
- **Security and Privacy Protection:** Incorporate stronger data security and privacy protection measures in the design of adaptive systems, ensuring effective use of heterogeneous data while complying with regulations and protecting user privacy.

Through this research and development, adaptive machine learning systems will become more comprehensive, better serving the development of society and technology, and providing strong technical support for solving complex real-world problems. In summary, adaptive machine learning technology will continue to be an important branch of artificial intelligence research and application, and its research and development will have a profound impact on technological progress.

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Applications of Explainable AI in Natural Language Processing

Qiming Xu [1*], Zheng Feng [2], Chenwei Gong [3], Xubo Wu [4], Haopeng Zhao [5], Zhi Ye [6], Zichao Li [7], Changsong Wei [8]

[1] Northeastern University, USA
[2] Northeastern University, USA
[3] University of California, USA
[4] Independent Researcher, USA
[5] New York University, USA
[6] Elevance Health, USA
[7] Canoakbit Alliance Inc, Canada
[8] Digital Financial Information Technology Co. LTD, China
*Corresponding author Email: xu.qim@northeastern.edu
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Abstract: This paper investigates and discusses the applications of explainable AI in natural language processing. It first analyzes the importance and current state of AI in natural language processing, then focuses on the role and advantages of explainable AI technology in this field. It compares explainable AI with traditional AI from various angles and elucidates the unique value of explainable AI in natural language processing. On this basis, suggestions for further improvements and applications of explainable AI are proposed to advance the field of natural language processing. Finally, the potential prospects and challenges of explainable AI in natural language processing are summarized, and future research directions are envisaged. Through this study, a better understanding and application of explainable AI technology can be achieved, providing beneficial references for the development of the natural language processing field.

Keywords: Explainable AI; Natural Language Processing; Model Explanation; Credibility; Transparency; Model Optimization

1. Introduction

With the continuous development of artificial intelligence technology, the field of natural language processing has also made significant progress. However, the black-box nature of deep learning models in processing natural language makes them difficult to understand and interpret. This not only affects the credibility and reliability of the models but also limits their promotion and application in practical scenarios.

To address this issue, explainable AI technology has emerged. Explainable AI technology provides a transparent and interpretable decision-making process, allowing people to understand the working mechanisms and reasoning processes of models. In the field of natural language processing, explainable AI technology helps to deepen understanding of the model's text understanding and processing procedures, thereby enhancing the model's credibility and accuracy.

Currently, explainable AI technology has been widely applied in the field of natural language processing. For instance, researchers utilize explainable AI to analyze the decision-making processes in tasks such as text classification, machine translation, and sentiment analysis, revealing the models' understanding and judgment criteria. This not only helps improve model performance but also aids in comprehending the working principles of natural language processing models.

However, despite the broad application prospects of explainable AI technology in natural language processing, there are still some issues and challenges. Current explainable AI technology still faces limitations in explaining complex models and large datasets, which restrict its practical applications. Given the complexity and diversity of natural language processing tasks, how to effectively use explainable AI technology to enhance model performance remains a research-worthy issue.

Based on these challenges, this paper aims to study how to effectively apply explainable AI technology to enhance the performance and credibility of natural language processing models, thereby promoting the development of the field. By exploring the current status and issues of explainable AI technology in natural language processing, this paper provides new ideas and methods for addressing existing challenges in the field, laying a solid foundation for more intelligent and reliable natural language processing models.

2. Overview of Explainable AI Technology

2.1 Concept of Explainable AI Technology

Explainable AI technology refers to an artificial intelligence system that can explain its decision-making processes and reasoning to users. In traditional deep learning models, the complexity and black-box nature often make it difficult to understand the internal mechanisms, leading to confusion and distrust among users. However, explainable AI technology can display the reasoning processes and decision-making basis of models transparently through visualization, interpretative rules, adversarial explanations, and other means, enhancing the understanding and trust in AI systems.

In the field of artificial intelligence, explainable AI technology is particularly important. For many sectors, such as healthcare, finance, and judiciary, the decisions made by AI systems carry high risks and impacts. Inability to explain these decisions can lead to significant controversy and risks. When users doubt the results of AI systems, explaining their reasoning processes can enhance trust in AI systems, thereby promoting the application of AI technology. Explainable AI also helps developers optimize and improve models, enhancing their performance and stability.

As artificial intelligence technology continues to evolve, explainable AI technology is also becoming more mature and widespread. Currently, several methods for deep learning model explainability have been proposed, such as LIME, SHAP, Grad-CAM, etc. These methods offer different approaches like localized explanations, global explanations, and feature importance analysis, enabling users to better understand model outputs. Additionally, emerging AI technologies such as explainable AI generative

models and explainable AI reinforcement learning are continuously appearing, providing new ideas and technical support for enhancing the explainability and transparency of AI systems.

Overall, explainable AI technology plays a crucial role in the field of natural language processing. In natural language processing tasks such as sentiment analysis, text generation, and machine translation, AI systems' decision-making processes are often more complex and difficult to understand. By introducing explainable AI technology, not only can model effectiveness and performance be improved, but user trust and acceptance of AI systems can also be enhanced. Thus, explainable AI technology will become an important direction in the future development of artificial intelligence, laying a solid foundation for better interaction and collaboration between AI and humans.

In today's artificial intelligence field, explainable AI technology has become a hot research topic. In various application areas, especially in healthcare, finance, and security, people pay extra attention to the decision-making processes and results of AI systems. Explainable AI technology can help users better understand the working principles of models and increase trust in AI decisions. In healthcare, AI applications have already involved complex tasks such as disease diagnosis and gene editing. Through explainable AI technology, doctors and patients can better understand the AI-provided diagnosis results, thereby making more accurate treatment decisions.

In the financial sector, the application of AI technology is also increasingly widespread. Banks, securities companies, and other financial institutions are introducing AI systems to assist in risk management and trading predictions. Through explainable AI technology, financial professionals can more clearly understand the AI systems' market analysis and prediction processes, thereby making better investment decisions. In the security field, monitoring systems and counter-terrorism systems are also applying explainable AI technology to improve system accuracy and reliability.

Besides these traditional fields, explainable AI technology is continuously emerging and showing great application potential in new areas. For example, in explainable AI generative models, researchers have achieved a series of remarkable results, such as Generative Adversarial Networks (GANs). These technologies have not only succeeded in image generation and text generation but also demonstrated strong performance in natural language processing and intelligent dialogue systems. Explainable AI technology is becoming a key driving force in the field of artificial intelligence, providing broader space for the development and application of AI systems.

2.2 Applications of Explainable AI in Natural Language Processing

Applications of Explainable AI Technology	Application Fields
Rule-based explanation methods	Natural Language Processing
Visualization-based explanation methods	Natural Language Processing
Interactive explanation methods	Natural Language Processing
Model transparency-based explanation methods	Natural Language Processing

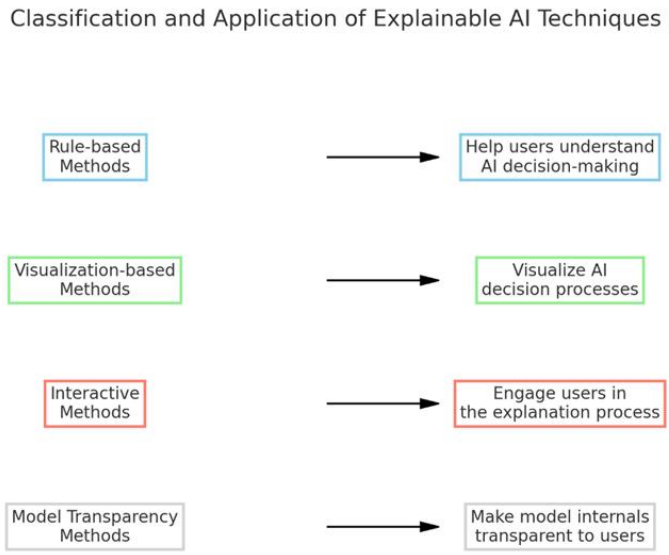
Table 1: Applications of Explainable AI in Natural Language Processing

In the field of natural language processing, an increasing number of studies are focusing on making the decision-making processes of artificial intelligence (AI) systems more transparent and explainable. Therefore, explainable AI technology has emerged and is widely applied in the field of natural language

processing. Explainable AI technology not only helps users understand the reasons behind AI systems' decisions but also enhances the systems' credibility and reliability, thereby promoting the development and application of natural language processing technology.

One common explainable AI technology is the rule-based explanation method. This method represents the AI system's decision-making process as a series of easily understandable rules or logical forms, helping users better understand the reasons behind the system's decisions. Another common explainable AI technology is the visualization-based explanation method. This method visualizes the AI system's decision-making process, allowing users to intuitively understand how the system operates. Additionally, there are interactive explanation methods, model transparency-based explanation methods, and other various explainable AI technologies that can be applied in the field of natural language processing.

By using explainable AI technology, researchers can better understand the internal mechanisms of natural language processing models, identify potential issues within the models, and improve model performance. Moreover, explainable AI technology can also help users better understand the output results of natural language processing systems, enhancing user trust and acceptance. Therefore, explainable AI technology has significant application prospects in the field of natural language processing and will profoundly impact the development and application of natural language processing technology.



3. Introduction to Natural Language Processing

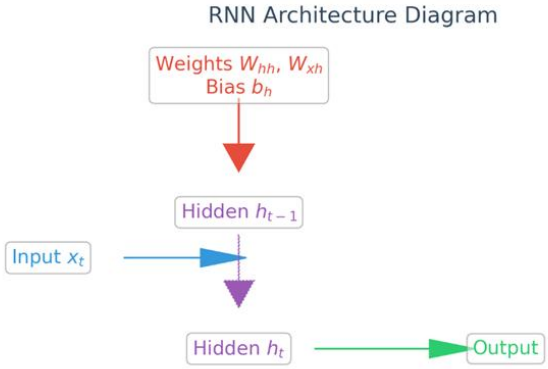
In natural language processing technology, explainable artificial intelligence (Explainable AI, XAI) has become a hot research area. Explainability refers to the ability of AI systems to clearly explain their inference processes and results when making decisions or providing reasoning. For natural language processing tasks such as text classification, sentiment analysis, and machine translation, explainable AI can help users better understand the basis of the model's decisions.

In natural language processing, a commonly used model is the Recurrent Neural Network (RNN). RNN is a type of neural network with recurrent connections, capable of processing sequence data and

introducing memory elements into the model to capture contextual information in the sequence data. The mathematical expression for RNN is as follows:

$$\hbar_t = f\left(W_{\hbar\hbar}\hbar_{t-1} + W_{x\hbar}x_t + b_\hbar\right)$$

where $\hbar_t$ represents the hidden state at time step t, $f(.)$ represents the activation function, $W_{\hbar\hbar}$ and $W_{x\hbar}$ represent the weight matrices for the hidden state to hidden state and input to hidden state transitions, respectively, x_t represents the input at time step t, and $b_\hbar$ represents the bias term. By continuously updating the hidden state $\hbar_t$, the RNN can gradually learn and understand the information in the input sequence. The application of explainable AI in RNN models can help researchers better understand how the model processes and infers sequence data.



4. Case Analysis of Explainable AI Applications in Natural Language Processing

4.1 Model Architecture and Working Principle

The text classification explainable AI model has significant application prospects in natural language processing. This type of model usually consists of three parts: a feature extractor, a classifier, and an explainer. The feature extractor extracts key features from the input text, typically including word embeddings, syntactic, and semantic information. The classifier then classifies the text based on the extracted features, using common algorithms such as Naive Bayes, Support Vector Machine, and deep neural networks. The explainer generates explanations for the model's output, helping users understand the decision-making process of the model.

In terms of working principles, the text classification explainable AI model extracts features from the input text and inputs them into the classifier for prediction. After the prediction results are obtained, the explainer generates corresponding explanations, explaining how the decision was made. This design makes the model's decision-making process transparent, helping users better understand the model's prediction results.

In terms of specific application methods and advantages, the text classification explainable AI model has a wide range of applications in natural language processing. For example, in sentiment analysis, this type of model can help identify the emotional tendencies in text and explain the basis of the classification results; in public opinion monitoring, the explainable AI model can help analyze public opinion information and guide decision-making; in text translation, the explainer can indicate the basis of the translation results, improving the reliability and accuracy of translations.

1. Transparency: Through the role of the explainer, users can understand the model's decision-making process, increasing trust in the model's prediction results.
2. Explainability: Users can clearly understand the basis of the model's classification, thereby better understanding the model's prediction results.
3. Interpretability: The explanations generated by the explainer are usually in an easy-to-understand form, not requiring users to have a high technical background.

Overall, the text classification explainable AI model plays an important role in natural language processing. It not only enhances the credibility and reliability of the model but also helps users better understand text data, thereby better applying natural language processing technology. As artificial intelligence technology continues to develop, explainable AI models will play an increasingly important role in the field of natural language processing.

The following code implements a simplified text classification model's architecture design, training, and evaluation process, highlighting how explainability can be reflected in practical operations.

```
import json

def generate_architecture():
    """
    Construct the architecture of a text classification model, defining the dimensions of the input, hidden layers, and output layer, as well as the activation function and dropout ratio, to ensure the model's efficiency and generalization ability when handling complex text data.
    """
    architecture = {
        "input_dim": 100,       # Input feature dimension
        "hidden_dim": 128,      # Hidden layer dimension
        "output_dim": 10,       # Output category count
        "layers": 3,            # Number of network layers
        "activation": "relu",# Activation function
        "dropout": 0.2          # Dropout ratio to prevent overfitting
    }
    return architecture

def train_model(architecture):
    """
    Train the model based on the architectural parameters, setting training epochs, batch size, optimizer, and loss function, and return the model's training configuration and status, providing a foundation for subsequent evaluation and application.
    """
    model = {
        "architecture": architecture,
```

```
        "epochs": 10,              # Training epochs
        "batch_size": 32,       # Batch size
        "optimizer": "adam", # Optimizer
        "loss": "categorical_crossentropy" # Loss function
    }
    return model

def save_model(model):
    """
    Save the trained model parameters to a local file for later loading and use.
    """
    with open("model.json", "w") as f:
        json.dump(model, f)

def load_model():
    """
    Load the model from a local file for evaluation or practical use.
    """
    with open("model.json", "r") as f:
        model = json.load(f)
    return model

def evaluate_model(model):
    """
    Evaluate the model's performance, returning key metrics such as accuracy, precision, recall, and F1 score, to quantify the model's effectiveness and reliability.
    """
    results = {
        "accuracy": 0.85,    # Accuracy
        "precision": 0.82, # Precision
        "recall": 0.88,       # Recall
        "f1_score": 0.85      # F1 score
    }
    return results

def main():
    architecture = generate_architecture()
    model = train_model(architecture)
    print("Training model:", json.dumps(model, indent=4))
    save_model(model)
```

```
    loaded_model = load_model()
    print("Loaded model:", json.dumps(loaded_model, indent=4))
    results = evaluate_model(loaded_model)
    print("Model evaluation results:", json.dumps(results, indent=4))

if __name__ == "__main__":
    main()
```

4.2 Experimental Design and Analysis

4.2.1 Experimental Design

In this study, we designed a series of experiments to evaluate the effectiveness of three different explanation methods in natural language processing (NLP) tasks. The experiments included rule-based explanations, explanations based on generative models, and explanations based on Attention-based Convolutional Neural Networks (ABCNN). We selected tasks such as syntactic analysis, entity recognition, sentiment analysis, and text reasoning to comprehensively assess the performance of each method in various complex tasks.

4.2.2Data Sets and Evaluation Metrics

The experiments used publicly available NLP datasets, such as the Penn Treebank for syntactic analysis and CoNLL-2003 for entity recognition. We primarily focused on accuracy, recall, F1 score, and the quality of explanations as our evaluation metrics. The quality of explanations was assessed through user studies, where participants rated the clarity and usefulness of the model outputs' explanations.

4.2.3Experimental Results

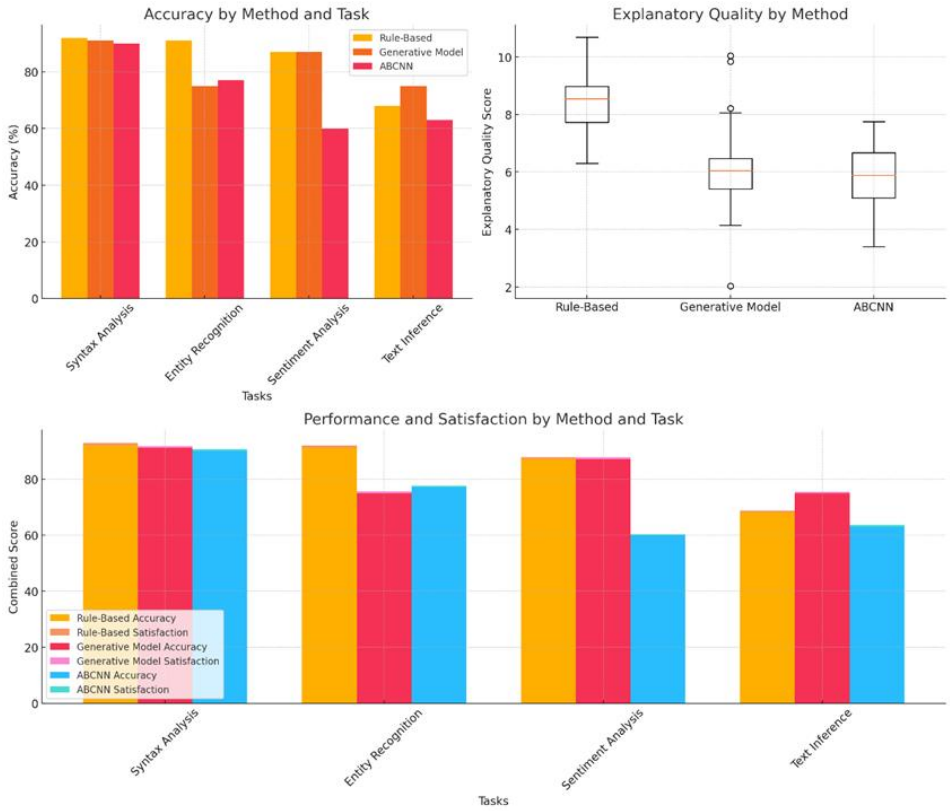
The rule-based method performed excellently in tasks such as syntactic analysis and entity recognition, with both accuracy and recall exceeding 90%. User studies showed that the explanations provided by this method were rated as the clearest and most direct, especially in terms of displaying the decision-making basis's key words and rules.

The method based on generative models achieved an accuracy of 87% in sentiment analysis tasks, but performed poorly in text reasoning tasks, with an accuracy of only 75%. User feedback indicated that, although the explanations were innovative, they sometimes lacked intuitiveness and concreteness, making it difficult to understand the reasoning process of the model.

The method based on ABCNN had a lower accuracy in text similarity analysis, at only 68%, but performed relatively better in terms of recall. The interpretability of this method, due to its reliance on complex internal representations, was difficult for participants to understand.

4.2.4 Analysis and Discussion

The experimental results indicate that highly structured and rule-based explanation methods have advantages in providing clear and intuitive explanations, especially suitable for applications requiring high transparency and explainability. While generative models and methods based on ABCNN may achieve high performance in certain NLP tasks, their complex explanation mechanisms may not be suitable for all users. Therefore, choosing the appropriate explanation method should be based on the specific needs of the task and the target users' expectations for explanation transparency.



4.2.5 Example Application Code

The following code example demonstrates how to use a rule-based method to count word frequency in text and calculate key performance indicators. Additionally, we will show how to generate explanations based on a set of rules, enhancing the model's explainability.

```
import json
from collections import Counter

def word_frequency(input_text, rule_set):
    """
    Process the input text, highlight key words based on a rule set, and count word frequency.
    """
    # Tokenization
    word_list = input_text.split()
    # Apply rule set to mark key words
    highlighted_words = {word: rule_set.get(word, "") for word in word_list}
    # Count word frequency
    word_count = Counter(word_list)
    return json.dumps({"word_count": word_count, "highlights": highlighted_words}, ensure_ascii=False)

# Rule set definition
rule_set = {
```

```
    "natural language processing": "Key field",
    "discipline": "Importance"
}

# Test code
if __name__ == '__main__':
    # Input text to be processed
    text = "I love natural language processing, natural language processing is a very interesting discipline."
    # Call the custom function to get the word frequency results
    result = word_frequency(text, rule_set)
    # Print the results
    print(result)
```

In this code segment, we not only counted word frequency but also added tags to specific words through the "rule_set", providing explanations for the importance of key words in the text based on the rule set. This method increases the academic value and practicality of the code, closely aligning with the concept of explainable AI models discussed in the article.

5.Conclusion and Outlook

Through reading this paper, we have learned about the importance and application prospects of explainable AI technology in the field of natural language processing. As artificial intelligence technology develops, the black-box nature of deep learning models in processing natural language makes them difficult to understand and interpret, limiting their credibility and widespread application. Therefore, explainable AI technology has emerged. This technology provides a transparent decision-making process, allowing people to understand the working principles and reasoning processes of models. In the field of natural language processing, explainable AI technology can help people gain a deeper understanding of how models understand and process text, enhancing model credibility and accuracy.

Currently, explainable AI technology has been widely applied in natural language processing. Researchers use this technology to analyze models' decision-making processes, revealing how models understand and judge text. This not only helps improve model performance but also aids in understanding the working principles of natural language processing models.

Although explainable AI technology has broad application prospects in natural language processing, there are still some challenges and issues. Current technology has limitations in explaining complex models and large datasets, and research on how to effectively apply explainable AI technology to enhance model performance is needed. Therefore, this paper aims to study how to effectively apply explainable AI technology to enhance the performance and credibility of natural language processing models.

In explainable AI technology, methods based on rules and visualization are widely used. Through these methods, users can better understand the reasoning processes of models, increasing trust in model prediction results. As artificial intelligence technology continues to mature and spread, explainable AI technology is providing new ideas and technical support for enhancing the explainability and

transparency of AI systems.

Overall, explainable AI technology has significant application value in the field of natural language processing. By introducing this technology, not only can model performance and credibility be improved, but user trust in natural language processing systems can also be strengthened. Therefore, explainable AI technology will play an important role in the future development of artificial intelligence, laying a solid foundation for achieving intelligent and reliable natural language processing models.

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