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The Commercialization Path of Large Language Models in Start-Ups

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Abstract: Recently, large language models (LLMs) have demonstrated strong performance in large-scale natural language processing tasks and have been widely adopted by many companies in the market. But in reality, for most start-ups, how to get their commercialization right is still a problem. This paper will analyze in detail the challenges posed to start-ups during the commercialization process of large language models, such as cost, economic constraints, compliance with relevant data protection laws, and adaptability to market demands, and provide solutions for technology integration, industry personalized design, and multiple commercialization models to help them successfully complete the commercialization process. In addition, it presents the benefits for start-ups in exploring new markets and globalization strategies. Finally, the paper summarizes the current commercialization status and future development trends of large language models and puts forward some perspectives and suggestions on the rational use of large language models to facilitate commercialization in start-ups.

Keywords: large language model; start-ups; commercialization path

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1. Introduction

As a core breakthrough in the field of artificial intelligence, large language models (LLMs) have been widely used in many areas such as text creation, translation, dialogue systems, and others. More and more start-ups choose technologies based on large language models to achieve corporate development and innovation. However, the commercialization of large language models is limited by the economic, technological, and market conditions faced by start-ups, which need to find efficient ways to utilize these models to achieve their goals [1-3]. This paper focuses on a detailed analysis of the current status and commercialization models of large language models in start-ups, especially providing solutions to key challenges faced in this process.

2. Big Language Model and Conceptual Framework for Start-Ups

2.1. Definition and Technical Background of Large Language Model

Large language models (LLMs) are natural language models trained using deep learning algorithms, which can learn linguistic rules and semantic information from large volumes of text data. The main architecture of large language models is based on the Transformer and uses a self-attention mechanism to process sequential data efficiently, extracting contextual information. Training large-scale language models typically involves two stages: pre-training and fine-tuning. During training, the model relies on large

amounts of unlabeled text data for unsupervised training to learn general linguistic patterns. Then, supervised training is conducted on specific tasks (such as language generation or question answering) to fine-tune the model’s performance. The research and application of large language models have had a great impact in many fields, such as text generation, machine translation, and sentiment analysis. Some prominent large language models, such as GPT and BERT, are also widely used in intelligent customer service, speech recognition, and translation [4].

Figure 1 shows the technical framework of the large language model, including the input embedding, Transformer processing layer, and output generation sections. By optimizing the information flow through the self-attention mechanism, the complex context information can be processed and the generation ability can be improved.

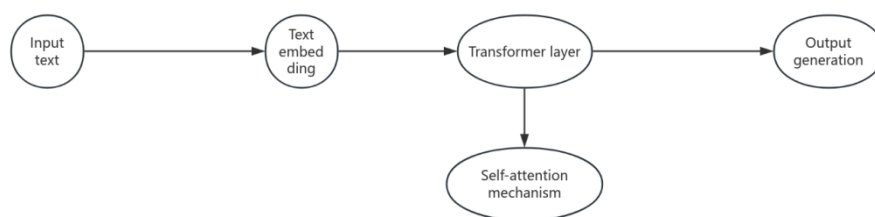


Figure 1. Large Language model technology framework.

2.2. Definition and Characteristics of Start-Ups

A start-up is a newly established business that relies on a particular product, technology, or service to open up a highly competitive market. The main characteristics of start-ups are the relative lack of funds and small number of employees, but they have great development potential. Despite limited funds, start-ups often exhibit strong flexibility by maintaining lean structures, quickly responding to market demand changes, and adopting agile R&D and iterative technological approaches to meet customer needs. Start-ups usually use angel investment, venture capital, and other financial instruments to obtain funding support. The biggest feature of start-ups is innovation. They generally use technology or business strategy innovation to increase their competitiveness in the market. However, due to market instability and financial pressure, startups are faced with greater risks and challenges, so they often adopt early-stage entrepreneurial approaches such as lean entrepreneurship, which involves rapidly launching products into the market and refining them based on customer feedback.

Table 1 shows five main characteristics of start-ups: high risk indicates that start-ups face large market and capital uncertainty; Rapid iteration refers to the frequent updating of products and services; innovation-driven highlights the role of technological advancement; limited resources point to constraints in capital and personnel; High flexibility refers to the ability of an enterprise to respond quickly to market changes.

Table 1. Characteristics of start-ups.

peculiarity	Description
High risk	Start-ups usually face insufficient funds and market uncertainty, and there is a high probability of failure.
Fast iteration	Startups often adopt agile development methods, quickly adjusting products and services, and continuously iterating and optimizing based on market feedback.
innovation-driven	Startups rely on technological innovation or unique business models to gain market share through differentiated competition and drive industry change.

Limited resources	Startups typically have less capital and staff and must rely on outside financing (e.g., angel investment, venture capital) to stay afloat and expand.
High flexibility	Start-ups have a relatively flat organizational structure, flexible decision-making processes, and are able to respond quickly to market changes and opportunities.

2.3. The Convergence of Large Language Models and Start-Ups

From the perspective of business application, large language models offer innovative advantages, cost effectiveness, and flexibility. Due to the relatively scarce resources of start-ups, manpower and financial resources are limited, and large language models can help them automate their work and reduce labor costs, such as service support and content generation. This gives start-ups more energy to focus on their core business. In addition, large-scale NLP systems offer a high degree of personalization. They can respond to the specific needs of start-ups and assist them in adopting new technologies to enhance their business competitiveness. Thanks to the adaptability of NLP, it can effectively respond to dynamic market demands, making it particularly suitable for supporting the rapid development of start-ups [5].

Figure 2 illustrates the three core benefits of a large language model for start-ups: technical support, cost effectiveness, and innovation. The proportions in the figure reflect the importance of these advantages in the start-up commercialization process, with technical support accounting for the largest share, indicating that the technical support provided by the big language model to the start-up is critical, while cost effectiveness and innovation capabilities play an important role in reducing costs and driving innovation.



Figure 2. Pie chart of meeting point between large language model and start-up enterprise.

3. Commercialization of Large Language Models in Start-Ups

3.1. Resource Consumption and Cost Bottleneck of Large Language Model

Although the potential of large language models cannot be underestimated, the high cost of hardware and computing power in the process of commercial landing has brought a lot of difficulties to entrepreneurial enterprises in the commercialization process. To achieve a high-performance model, a large number of advanced computing units such as GPUs are required, which leads to significant procurement and maintenance costs as well as substantial resource consumption. The cost of training a large language model can be estimated by the following formula:

$$C_{train} = N_{GPUs} \times C_{GPUperhour} \times T_{traininghours} \tag{1}$$

Where, C_{train} denotes the training cost, N_{GPUs} is the number of Gpus used, $C_{GPUperhour}$ is the cost of Gpus per hour, $T_{traininghours}$ and is the total time required for training. The training process often takes weeks or even months, making hardware investment very expensive for startups.

In addition, the working logic of the large language model (i.e. the actual application process) also generates a lot of computing power. When a startup deploys the model in a real production environment, responding to each user query requires significantly more computational power than a single forward training pass. With the increase in the number of users, the consumption of servers and computing power will increase significantly, increasing the operating costs of enterprises. When startups face resource constraints, they must find ways to deploy and optimize existing resources wisely. Excessive computing and energy consumption complicate the commercialization process and threaten sustainable development [6].

3.2. Data Privacy and Security Compliance Issues

When the large language model is commercialized, how to protect the user's data security and privacy has become the first thing for start-ups to consider. In order for it to work best, it often requires a large amount of information data containing users and sensitive information to be trained. Therefore, ensuring the security of this kind of information and using this information in accordance with relevant laws has become a problem that start-ups must solve before they want to promote it on a large scale. All countries have strict legal control over information security. For example, strict data protection regulations adopted in Europe and domestic laws in China impose stringent requirements on legal entities regarding the collection, storage, and use of personal information. Therefore, startups that intend to use AI assistants must ensure their data collection and processing methods comply with relevant laws and regulations to avoid legal challenges, significant fines, and reputational damage. Although technically encryption, anonymization, and other methods can ensure data security, startups often lack sufficient resources and experience. Balancing data privacy protection with business needs has become a major challenge for their operations.

Table 2 lists the main data privacy and security issues that startups face when applying large language models, including ensuring compliance with laws such as GDPR and PIPL, ensuring secure data storage and transmission, preventing data breaches, and protecting personal information through anonymization techniques to avoid legal risks and security threats.

Table 2. Data Privacy and security compliance issues.

problem	Description
Legal compliance	Startups must comply with national data protection laws such as GDPR, PIPL, etc.
Data security	The security of data storage, processing and transmission must be ensured to prevent data leakage.
Data anonymization processing	Use technical means to ensure that personal information is not leaked and to maintain data anonymity

3.3. Insufficient Market Suitability of Application Scenarios

Although the large language model has great potential in the large-scale application industry, many start-ups will face the bottleneck of not adapting to the market in actual operation. To maximize the capabilities of large language models, start-ups need to carefully select or tailor their application environments to ensure greater applicability and effectiveness. However, many startups do not decide the direction of their products according to the market demand, but blindly apply the large language model to the unsuitable market, which leads to the incompatibility of technology use and market demand. Although large language models demonstrate advantages in broad areas such as automated customer support, copywriting, and sentiment analysis, their benefits are less pronounced in specialized industries or segments. For example, in a highly technical field such as law

or medical industry, it is difficult to accurately predict some conclusions due to the lack of industry knowledge background of the model. If a startup does not understand actual market needs, its technology development may deviate from demand, resulting in misaligned applications and wasted resources [7].

4. Exploration and Evolution of Commercialization Paths of Large Language Models

4.1. Technology Integration and Industry Customization

For large language models, commercial landing is not only the basic ability, but also the use of technology overlay industry personalized customization to achieve the needs of the actual market. In the innovation scenario of a new enterprise, adapting the big language model based on industry-specific data sources allows the big language model to accurately provide solutions to business problems. For example, in the financial industry, large language models are customized based on domain-specific financial terminologies and data characteristics to provide users with improved risk prediction and market decision-making. Technology connectivity is the seamless docking of large natural language processing platforms with existing technology platforms (such as databases, cloud servers, API connections, etc.) to ensure high efficiency and compatibility with other business processes. With this technological connection, startups can not only enhance the model's capabilities, but also ensure its efficiency and scalability.

Figure 3 shows the technology integration framework for large language models and industry customization, including collaboration between industry data sources, custom models, and other technical systems. The framework illustrates how to optimize the application of large language models in specific industries through the combination of data sources, models and technical systems.

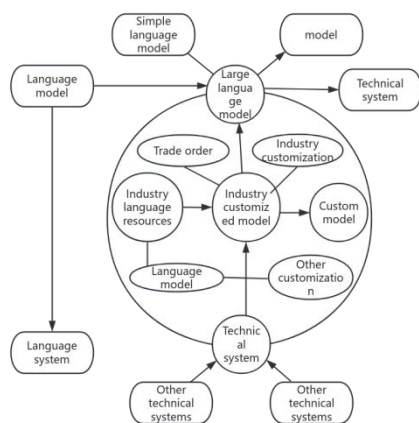


Figure 3. Large language model technology integration framework.

4.2. Diversified Business Model: SAAS and Platformization

From the perspective of large language models, their commercialization path is not limited to a single application type. Most startups adopt various commercialization strategies to expand their business scale. The most widely known approaches are SaaS (Software as a Service) and platform-based models that offer SaaS services. SaaS offers customers the ability to use a large language model online without requiring them to build or maintain a hardware platform. This approach reduces the technical burden on start-ups and makes it easier to attract more users as quickly as possible. For example, startups can use API services provided by large-scale AI models to allow customers to access the model's capabilities through interfaces, completing natural language tasks such as text generation, voice recognition, or machine translation. In this way, innovative enterprises can

expand their business areas and obtain continuous and stable income through subscription fees. In addition, the SaaS (Software as a service) model has a flexible price, which can be used by small and medium-sized enterprises, thus increasing the customer's dependence on enterprise services. The platform model is a sustainable profit model. Startups can develop a service environment featuring numerous AI assistants that combine various functions and use cases to address a wide range of problems. The biggest advantage of this model is that it can attract a large number of users and developers, and build a good ecosystem [8].

Table 3. Comparison of SaaS and platform-based business models.

mode	Description	advantage	Challenge
SaaS	Provide API services based on large language models, charging on demand.	Flexible pricing, low threshold, wide market coverage.	Requires powerful cloud computing resources and relies on continued subscriber growth.
platforming	Build a service platform to integrate different application scenarios.	It can provide diversified services and form an ecosystem.	The cost of platform construction is high, and the ecosystem of developers and users needs to be maintained.

Table 3 compares the two business models of SaaS and platformization. By providing API services based on a large language model, the SaaS model has flexible pricing, low threshold and wide market coverage, but it also faces the challenge of high dependence on cloud computing resources and user growth. The platform model can form a service ecosystem by integrating a variety of application scenarios, but its platform construction and maintenance costs are high.

4.3. Market Expansion and Globalization Strategy

Start-ups should not only have a plan to expand in the domestic market, promoting large language models, but also actively explore overseas channels and implement internationalization strategies. Given the advanced natural language processing capabilities of the large language model, it has an extremely strong demand in various foreign regional markets, including multilingual support, cross-cultural, text generation, and intelligent customer service worldwide. If a start-up wants to go to sea, it needs to understand the market characteristics and laws and regulations of overseas countries. Start-ups should pay attention to the local and transnational issues in the marketing of new products. For non-English speaking markets, especially in Asia, Europe and other regions with large differences, it is necessary to localize and personalize its large natural language processing platform at the beginning of its establishment to ensure the provision of global and diversified intelligent services. As data protection laws differ in each country and region, local laws and regulations need to be followed when expanding internationally. By adopting a global strategy, start-ups are able to expand their sales territory to a larger extent and strengthen their technological capabilities by leveraging the strengths of technologies and markets in different regions. For example, in developed markets such as the United States, startups can improve their products by leveraging advanced AI markets and mature technological infrastructures. In Asia-Pacific countries like China, adopting local sales strategies is necessary to meet rapidly growing demand.

Figure 4 shows the market expansion framework for startups implementing globalization strategies. Core elements such as localization, data compliance and multilingual support are included in the chart to help startups understand how to effectively scale their

markets globally and adapt their technology and services to meet the needs of different regions.



Figure 4. Framework of global strategic market expansion.

5. Conclusion

LLMs provide startups with powerful capabilities to improve work efficiency, reduce costs, and stimulate innovation. However, in the process of commercializing their applications, they still face significant challenges related to resource constraints, data privacy law compliance, and market adaptation. These challenges are inevitable, but by integrating technology into their products and customizing it for each industry, startups have the potential to overcome these challenges and promote the broader adoption of large language models (LLMs). As the technology matures and the market evolves, large-scale adoption of these machine learning-based systems will become increasingly common.

References

1. L. Yun, "Analyzing Credit Risk Management in the Digital Age: Challenges and Solutions," *Econ. Manag. Innov.*, vol. 2, no. 2, pp. 81–92, Apr. 2025, doi: 10.71222/ps8sw070.
2. S. Yang, "The Impact of Continuous Integration and Continuous Delivery on Software Development Efficiency," *J. Comput. Signal Syst. Res.*, vol. 2, no. 3, pp. 59–68, Apr. 2025, doi: 10.71222/pzvfqm21.
3. F. Gao, "The Role of Data Analytics in Enhancing Digital Platform User Engagement and Retention," *J. Media Journal. Commun. Stud.*, vol. 1, no. 1, pp. 10–17, Apr. 2025, doi: 10.71222/z27xzp64.
4. M. Bourdin et al., "Exploring the applications of natural language processing and language models for production, planning, and control activities of SMEs in industry 4.0: a systematic literature review," *J. Intell. Manuf.*, 2024, pp. 1–21, doi: 10.1007/s10845-024-02509-w.
5. J. Lee, W. Jung, and S. Baek, "In-house knowledge management using a large language model: focusing on technical specification documents review," *Appl. Sci.*, vol. 14, no. 5, p. 2096, 2024, doi: 10.3390/app14052096.
6. A. A. Linkon et al., "Advancements and applications of generative artificial intelligence and large language models on business management: A comprehensive review," *J. Comput. Sci. Technol. Stud.*, vol. 6, no. 1, pp. 225–232, 2024.
7. M. Banka et al., "Understanding corporate concerns. Barriers and challenges in corporate–start-up collaboration," *J. Open Innov. Technol. Mark. Complex.*, vol. 10, no. 4, p. 100388, 2024, doi: 10.1016/j.joitmc.2024.100388.
8. S. Rubach, "Leveraging a dual Organisation-Development (OD) Process: a longitudinal study of a start-up's transformation in the energy sector," *Syst. Pract. Action Res.*, vol. 37, no. 5, pp. 585–600, 2024, doi: 10.1007/s11213-024-09695-x.

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