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Optimizing Game Conversion Rates and Market Response Strategies Based on Data Analysis

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Abstract: With the rapid development of mobile Internet technologies, the gaming market has become a significant sector within the digital entertainment industry. In such a challenging market environment, game developers increasingly focus on data analysis to optimize conversion rates and reduce market response time. Game conversion rate refers to the proportion of users who engage with a game and complete specific actions such as registration and payment. It is a key metric for measuring game success. Market response includes factors such as advertising effectiveness and user feedback. By analyzing game conversion rates and market response through data, game developers can continuously adjust user experience and market response strategies to maximize overall revenue. This paper discusses the main issues in applying data analysis to optimize game conversion rates and market responses and offers solutions.

Keywords: game conversion rate; market response; personalized recommendation

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1. Introduction

With the development of network technology, competition within the gaming industry has become increasingly intense. To stand out in the market, game developers must constantly improve the conversion rate and market response. Conversion rate refers to the percentage of new players who successfully register and complete payment after trying the game. It is a quantifiable measure that reflects the game's development and operational adjustments before release, based on ongoing data analysis. Market response refers to the consumers who enter the game due to the influence of advertising or marketing, with their experiences reflected in different forms. By applying data analysis tools, developers can better understand the diverse needs of players, adjust the game's design and planning structure, and enhance player experience and purchasing desire. However, in practice, challenges such as data collection and privacy protection remain. Thus, effectively using data analysis to improve conversion rates and market response has become a critical issue in the industry.

2. Importance of Optimizing Game Conversion Rates and Market Response Based on Data Analysis

For the gaming industry, improving conversion rates and market responsiveness are key to ensuring long-term gaming revenue. Conversion rate refers to the entire process of users from initial exposure to the game to registration and consumption, while market response refers to the impact of marketing activities on player behavior. Through data analysis, developers can accurately understand players' behavior habits and preferences,

thereby optimizing game content and marketing strategies [1]. The core of improving conversion rates lies in reducing user churn and increasing paid conversion rates. By analyzing the reasons for player churn, developers can adjust game difficulty, optimize task design, improve reward mechanisms, etc., to ensure that players can find fun in the game and are willing to invest in it for the long term. Data analysis helps identify players' pain points, enabling precise content adjustments, increased retention rates, and paid conversions. In terms of market response, data analysis can monitor advertising performance, user feedback, and activity engagement in real time, helping developers adjust their promotional strategies in a timely manner and maximize advertising return on investment (ROI). By understanding which advertising formats and channels are more effective, developers can allocate resources more efficiently, optimize advertising placement, and ensure maximum effectiveness in acquiring new users [2]. In short, by optimizing the conversion rate and market response of games through data-driven analysis, developers can enhance the competitiveness of games, strengthen their market position, and ensure the long-term success and profitability of games.

2. Specific Applications of Data Analysis in Game Optimization and Market Response

2.1. Player Behavior Analysis and Personalized Recommendations

By analyzing and observing users' in-game behavior, we can understand their preferences, needs, and potential churn risks. By analyzing users' game time, payment information, and clearance records, we can understand potential customers who may leave and their specific reasons for leaving. For customers who may experience churn, companies can identify specific issues by analyzing their in-game communication, game consumption patterns, and other factors, such as the difficulty of the game, lack of attractive rewards, or lack of appeal in certain game features. By analyzing user preferences such as completed tasks, selection of virtual items, and consumption behavior, developers can clearly identify the game elements or parts that players prefer the most [1]. This data can be used to implement personalized recommendation systems to increase player engagement and conversion rates. The formula for personalized recommendation is as follows:

$$r(u, i) = \frac{\sum_{\theta \in N(u)} w(u, \theta) \cdot r(\theta, i)}{\sum_{\theta \in N(u)} |w(u, \theta)|} \quad (1)$$

$r(u, i)$ represents the predicted rating of item i by player u ; $N(u)$ is the set of other players similar to player u ; $w(u, v)$ represents the similarity between players u and v (usually calculated by cosine similarity); $r(u, i)$ is the actual rating of item i by player v . Using this formula, the recommendation system can suggest items or game content that players are likely to be interested in.

2.2. Real-Time Monitoring and Dynamic Adjustment

Real-time monitoring of players' behavior in the game and dynamically adjusting game content can make players more engaged and reduce churn rates. By real-time analysis of players' game duration, task completion status, interaction frequency, payment behavior, and other data, changes in players' interests can be revealed. Developers use a data monitoring platform to continuously collect game data and analyze it, identifying problems in the data in a timely manner and making timely adjustments. The game churn rate refers to the ratio of all inactive players to the total number of players over a period of time. Based on real-time monitoring data, developers can dynamically adjust game content to enhance player engagement and retention rates. Dynamic adjustment is not limited to difficulty balance and level design within the game, but also includes task structure, reward mechanism, social interaction, and other aspects. By analyzing players' activity status and in-game behavior in real-time, developers can discover that a certain stage or activity has a high churn rate, and adjust the difficulty of the link or activity, the design of rewards, and the addition of reminder settings to reduce its churn rate [3]. The churn rate formula is:

$$\text{churn rate} = \frac{\text{number of churned players}}{\text{total number of players}} \times 100\% \quad (2)$$

2.3. Market Response and Advertising Optimization

The optimization of market response relies on real-time data analysis of player behavior and advertising effectiveness. Through data analysis, developers can evaluate the effectiveness of advertising campaigns, such as click-through rates, conversion rates, customer retention rates, and other core data. Based on real-time data analysis of advertising effectiveness, developers can further optimize advertising content and target audiences, ensuring that advertisements can effectively reach potential paying players [4]. Accurate audience analysis helps developers develop personalized advertising strategies, thereby improving the effectiveness of advertising. By analyzing data from different advertising channels, developers can understand which media or methods have higher player appeal, thereby optimizing their advertising publishing decisions. Meanwhile, by analyzing players' game behavior, consumption habits, and payment history, developers can predict which players may make payments in the future. Data analysis can also help developers identify potential high paying players, providing them with personalized discounts or rewards to incentivize them to make more purchases. When players reach a certain milestone, developers can incentivize them to make purchases through discounts, gift packs, or limited-time events. Through this approach, developers can maximize players' payment potential and increase revenue. The formula is as follows:

$$ROI = \frac{\text{advertising revenue} - \text{advertising cost}}{\text{advertising cost}} \times 100\% \quad (3)$$

3. Problems Facing Game Conversion Rates and Market Response

3.1. Missing Information in Data Collection

Data collection is fundamental to optimizing game conversion rates and market responses. However, during the process, gaps in the data can lead to misleading conclusions, causing developers to miss key player behavior or fail to record important data. Tracking may be disrupted during gameplay due to system issues or user settings, and network failures might result in incomplete data. Additionally, if individual data is not collected due to inadequate monitoring methods, it could negatively impact player conversion. Beyond technical barriers, another challenge in data collection is coverage. For example, players' cross-platform behaviors or multiple device usage may not be effectively integrated, causing difficulties in analyzing cross-platform or multi-device usage behavior. Without timely collection of indirect behaviors such as social interactions, comments, or feedback, it also limits understanding of overall player behavior, affecting game optimization and market response strategies.

3.2. Abstract Analytical Results without Practical Implementation

While data analysis can reveal valuable insights into player behavior, game content, and market responses, conclusions often lack clarity and actionable guidance. Developers may possess extensive datasets and identify trends, yet still lack clear, practical recommendations. For example, analyzing high churn levels in certain game stages may lead to the conclusion that "the level design needs optimization", but how should the level be adjusted? Should the difficulty be increased or decreased, or should rewards be improved? These types of questions cannot be answered directly from the analysis reports. A similar situation occurs in market response analysis, where developers may discover low conversion rates for certain ads but may not know whether the issue is with the ad strategy itself, the messaging, the target audience, or the advertising channel.

3.3. Data Collection without Explicit User Consent

With increasing concerns about data privacy and the introduction of privacy regulations such as the EU's GDPR, failure to obtain explicit user consent for data collection has

become a significant issue. If developers do not inform players about the purpose and usage of their data or fail to obtain explicit consent, they risk facing legal and ethical challenges. Players may lose trust in the data collection process, which can lead to user dissatisfaction and reduced engagement. In extreme cases, players may leave the game or leave negative reviews, damaging the game's reputation. In addition, disorderly data collection can result in regulatory penalties and economic losses. For developers, collecting and using player data transparently and legally has become a challenge. As shown in the Table 1 below:

Table 1. Major Issues and Their Impact on Game Conversion Rates and Market Response.

Issue	Description	Impact
Data Collection Omissions	Due to technical limitations, device issues, or privacy settings, some player behavior data is not fully collected.	Incomplete data, biased analysis results, affecting the accuracy of game design and market response strategies.
Abstract Analysis Results	Data analysis results lack specific implementation directions and fail to provide actionable suggestions.	Analysis cannot be converted into effective optimization actions, reducing the practical value of data analysis and affecting the game optimization process.
Lack of User Consent for Data Collection	Player data was collected without explicit consent or in violation of privacy protection regulations.	Potential legal risks, player churn, and damage to brand reputation, impacting market response and player conversion rates.

4. Strategies for Optimizing Game Conversion Rates and Market Response

4.1. Enhance Data Capture Channels to Track User Changes

Data capture is essential for optimizing game conversion rates and market responses, as it enables developers to monitor player behavior and respond in real time. Enhancing data capture channels ensures that developers can timely acquire comprehensive player data. By strengthening data capture channels, developers can track player behavior in real time. In addition to regular in-game behavior tracking, monitoring of player social interactions, devices, and cross-platform behavior can also be added. Furthermore, increasing the frequency of data collection and reducing latency helps in real-time understanding of player changes [5]. Data capture channels can include a combination of game clients, social media interactions, user feedback, ad clicks, and more, ensuring comprehensive tracking of player behavior changes. As shown in the framework flowchart below (Figure 1).

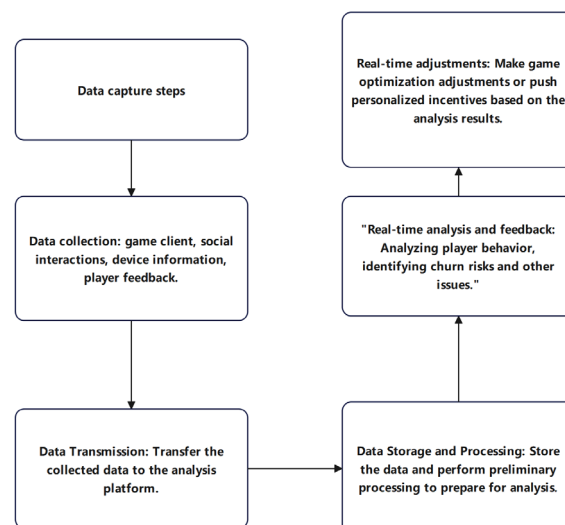


Figure 1. Data Capture Flowchart Steps.

For example, Tencent's most popular mobile multiplayer battle game, Honor of Kings, has enhanced its data collection methods to improve the game's conversion and retention rates. The research team collects player behavior data and monitors player changes in real time. In addition to collecting data from the game itself, the project team also gathers player behavior data from other channels. For each match, detailed player actions are recorded, including kills, deaths, assists, tactical actions, and more. After each match, the data for every participant is recorded and analyzed. The team also collects all in-game social interactions, including friend invitations, team formations, and chat records, to better understand changes in player relationships. Additionally, Data on players' phone models, operating system versions, and network stability is collected to identify technical flaws and improve game performance. Regular player surveys and feedback are used to assess satisfaction with the game content and overall experience, complemented by quantitative data for in-depth research. To monitor players' offline behavior, the game team has integrated offline alerts into the data collection system. To detect signs of player attrition, a warning mechanism has been introduced. By tracking player activity (e.g., login frequency, gameplay duration, participation in battles), if a player becomes inactive for a period or shows signs of fatigue in the game (such as reduced competitive actions or less interaction), the system flags them as at risk of churning.

4.2. Use of Machine Learning Algorithms for Gradual Data Optimization

Through machine learning technology, game developers can mine data from users' behavior paths to understand their actions, segment players into different groups based on their behavior, and then carry out targeted optimizations. First, clustering algorithms (e.g., K-means) are used to divide players into high, medium, and low engagement groups. Then, classification algorithms (e.g., decision trees) are applied to predict the behavior trends of each group, identify churn risks, and assess payment potential. For high-engagement players, advanced items or skins are recommended; for medium-engagement players, limited-time activities and rewards are offered; for low-engagement players, beginner guides are optimized, and game difficulty is reduced to decrease churn. Progressive user segmentation helps developers deliver more precise content recommendations, thereby improving conversion rates, player retention, and revenue. As shown in Figure 2:

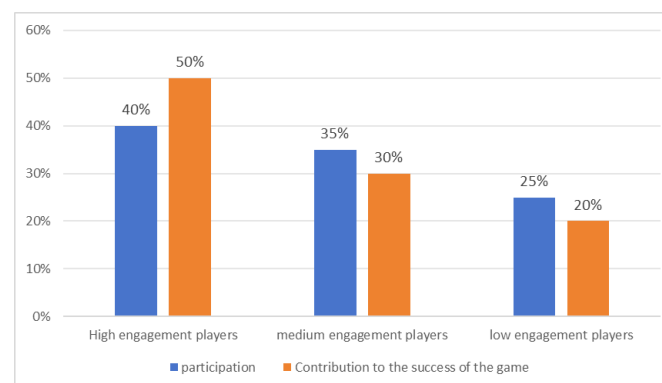


Figure 2. Player Engagement Distribution and Their Contribution to Game Success.

For example, a certain mobile game, in order to improve its conversion rate, used machine learning to analyze users' behavioral history, including consumption habits, shopping time, preferences, and other information. As a result, it successfully classified several types of users, such as frequent paying users, occasional paying users, and never paying users, and developed personalized paid recommendations for different categories. For players who frequently make purchases, the system recommends more high-value products or packages and offers them special rewards, while for occasional spenders, it provides time-limited offers and discounts to encourage further purchases, offer them

some expiration dates and discounts to attract them to buy more. Through this type of segmented recommendation method, the game can continuously adjust the form of advertising promotion and discount strategies according to the different attributes of users, improving its own payment rate. As the system continues to learn and optimize, the game's paid recommendations become more and more accurate, improving users' payment behavior.

4.3. Strengthen Data Privacy Protection and Add Privacy Transparency Policies

With the increasing enforcement of global data privacy laws such as the GDPR, game developers must ensure that the data they collect and use of player information is public, legal, and compliant. Improving the protection mechanism for player privacy can ensure the security of player data while gaining their trust, reducing the risk of player data leakage, and making the player experience more satisfactory. The privacy transparency policy should clearly inform players of which data is collected, how it is used, and how to protect their privacy. Providing transparent privacy settings can make players feel more at ease, thereby enhancing user trust and long-term retention.

For example, Fortnite is a globally popular game developed by Epic Games. In order to ensure that the game complies with European GDPR and data privacy regulations in other regions, the privacy policy of the game details the purpose and methods of data collection, including collecting players' hardware information, consumption data, game progress, etc., providing customized gaming experience, and improving game quality. At the same time, Epic Games declares to players that the collected data is only used for game improvement and promises not to use user information for other purposes. Moreover, Fortnite provides control over users' private settings, allowing players to view and change their personal profiles at any time. Players can choose to opt out of data collection or delete their stored data, which enhances their sense of trust. By adopting this privacy protection strategy, the game has avoided potential legal risks, gained player trust, increased player retention and game activity, and promoted the development of the gaming market.

5. Conclusion

With the continuous development of the gaming industry and increasingly fierce market competition, optimization strategies based on data analysis have played a crucial role in improving game conversion rates and market response. Game developers rely on data analysis to analyze players' operational behavior in a targeted manner, increasing player engagement and consumption conversion rates. However, data collection and privacy protection remain one of the challenges faced by developers. How to collect and use data while protecting player privacy is an important issue currently faced by game developers. In the future, with the continuous advancement of technology, data analysis will play a greater role in the gaming industry, helping game developers better understand the needs of players and achieve the success and long-term development of games.

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