

Analyzing the Impact of Scientific and Technological Innovations on Financial Performance: A PCA-Based Study of an AI Healthcare Startup

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Abstract – This study explores the relationship between scientific and technological innovations and the financial performance of startups, focusing on a representative Artificial Intelligence (AI) healthcare startup. Utilizing Principal Component Analysis (PCA), the research aims to dissect the complex interplay between innovation-related metrics—such as R&D spending, patent counts, and technology adoption rates—and financial outcomes like revenue growth, profitability, and market share. The PCA methodology enabled the reduction of high-dimensional data into PCA, which clearly illustrates how several dimensions of innovation impact financial metrics. The approach used in the investigation helps people comprehend more thoroughly how development benefits business viability in several different manners and provides startups with practical advice to use when preparing their revolutionary investments. This paper aims to assist the ecosystem of startup consumers (including investors, business owners, and policymakers) in making better decisions that balance technological progress with economic objectives.

Keywords – Artificial Intelligence, Principal Component Analysis, Finance, Policy-Makers, Business Models.

I. INTRODUCTION

Startups are essential in promoting innovation and economic development in modern competitive economic performance, especially in the information technology industry. These businesses introduce novel, cutting-edge goods and services to the market and perform a significant part in strengthening economic growth while producing employment. However, numerous challenges from invention to Financial Sustainability (FS), render this trend of startup success and survival extremely complicated and driven by several variables.

A startup's competitive edge originates due to its revolutionary approach. Transforming economic sceneries is not limited to the availability of new products or innovations; it additionally involves creative company models and operations. Maintaining up with the most recent technologies is essential for new businesses. It allows them to differentiate from other more established businesses and respond rapidly to evolving demands from the marketplace.

FS encompasses more than merely making money for business people. Developing an environmentally friendly business model is crucial for supporting development and securing the financial capital required to stimulate perpetual

innovation. In order to draw in investors and sustain scientific and technological tasks, it is essential to demonstrate that novel concepts consistently generate economic benefits.

Innovation is significant, but it's challenging to determine a number on how much of an impact it has on businesses' bottom lines. When evaluating the short-term value of revolutionary technology or novel business methods, conventional financial standards could fall short. It has become challenging to determine the direct financial impact of technology due to the delayed operation, which arises when the benefits of technology are not realized until after the primary investments have been achieved.

The research proposal will employ an enormous data set from an artificial intelligence medical services startup to examine the connections between investment in innovation, the total amount of trademarks granted, and the progression of technology implementation, as well as financial findings like sales growth, economic viability, and stake in the market. Through the application of Principal Component Analysis (PCA), this study will distill complex, high-dimensional data into PCA that reflect the most significant relationships between innovation activities and financial results, offering strategic insights that could guide future investments and business strategy formulation in technology-driven startups.

This study aims to bridge the gap in understanding how scientific and technological innovations influence the financial performance of startups.

By Employing Advanced Analytical Techniques, Such as PCA, This Research Seeks to

- (a) Identify and quantify the key factors of innovation that significantly impact financial metrics.
- (b) Develop a model that can predict financial performance based on varying levels of investment in innovation.
- (c) Demonstrate proof that investment ideas in innovation-driven companies perform.

The research paper has been organized in the following order: the literature survey is given in part 2, the approach is shown in section 3, the results and analysis will be displayed in section 4, and the work is concluded in section 5.

II. LITERATURE REVIEW

In their research on renewable energy developments and their financial impact, [1-2] concentrates on China's production industry. Through the prism of sustainability and competitive advantage, their approach evaluates numerous ways green manufacturing and goods enhancements improve economic growth. Green product development enhances financial performance, based on their empirical examination of 642 manufacturing companies, whereas innovation in environmentally friendly processes boosts it. Significant conclusions for sustainable development strategy may be derived from this variation, which emphasizes the different methods numerous green innovations impact financial outcomes.

In their research on the information technology industry in the United States, [2-3] examined how CSR and economic development relate. They demonstrate that using precise computational methods, online businesses' efforts in environmentally conscious operations run together with greater profitability and revenue. This study throws uncertainty on previous assumptions on the impact of CSR on the value of stocks, especially Tobin's Q. It claims that, while CSR might improve sales and profits directly, the effect on market valuation remains ambiguous. The research conducted here adds to the ongoing discussion on how information technology companies ought to weigh the objectives of their stockholders with their responsibilities to other stakeholders by emphasizing the financial viability of CSR.

In the review of green innovation's impact on China's economic development, [4-5] emphasize the nation's northwest and central regions. Researchers utilize an empirical approach to demonstrate that innovations in natural environments, supported by research-motivated educational systems, suggestively improve environmental guidelines and reduce pollution. It is important to note that their study indicates that sustainable financing techniques are likely to reduce energy conservation excess investment and increase the cost-effectiveness of renewable energy investments. Many of the advantageous impacts of sustainable practices regulations on local economies have been demonstrated here by the complex relationship between innovative environmental and economic strategies.

Sustainable development is the primary focus of [6-7] research into how technological innovation impacts the success of startups. Advances in technology significantly improve a business's credibility and economic performance, based on their study of 204 startups that employed multilevel regression techniques. The relationship shows how ethical behavior may boost profitability and provides weight to the theory that startup firms could reap advantages from adopting FS policies.

The effect of critical economic developments on the financial results of SMEs during the COVID-19 pandemic was studied by [8-9]. Researchers determined that in order to overcome the financial challenges imposed by the global epidemic, sales and manufacturing enhancements were essential based on their study, which was based on a sample of Lebanese SMEs. On the other hand, throughout this time, improvements to both goods and business procedures did not significantly impact financial results. By emphasizing the function of particular types of innovation in securing business succession and profitability, this difference highlights the vital role of adaptive innovation techniques in dealing with emergencies.

Addressing Islamic banks in Bangladesh, [10-11] examine how delighted consumers are with the service they get. Their investigation indicates a complex relationship between trust and client loyalty, with tangibles, understanding, reliability, and availability all positively impacting loyalty. Also, additional factors may have a more significant impact on financial outcomes than loyalty to customers, as there is a lack of association between the two and the financial performance of these businesses.

III. METHODOLOGY

Data Collection

This study's empirical basis is based on significant data collected from InnovaTech Ltd., a Chinese business highlighting artificial intelligence technology for use in healthcare. An in-depth examination of the functional and economic trends within the company was accomplished via the data-collecting process, requiring an in-depth review of core and outward sources [12-13].

*Key Sources Include**Internal Company Records*

The business provided substantial internal records. Include budgets, study and development expenditure data, and schedules for introducing novel products. These reports provide an overview of the three-year budgets and outcomes of the business's innovation-driven approach.

Public Filings and Reports

It is simple to find publicly available information such as financial statements and declarations with the SEC. Reliable financial data on the company's market behaviors, requests for patents, and adoption of technology statuses can be attained from these records.

Intellectual Property Databases

Databases on a national and international level that track intellectual filings, providing data on the number and standard of patents filed.

Nature and Size of the Dataset

Approximately 600 distinct facts have been collected periodically from 2020 to 2023 to make up the dataset employed for the purpose of this study. In order to perform an accurate historical study, every data point includes quantitative indicators of innovative activity and financial results. Due to the dataset's broad coverage of all innovation phases and financial periods, it is feasible to accurately assess the time-varying connection between investment in research and development and economic benefits.

*Variables of Interest**Innovation Metrics*

Innovation. is quantified using the following metrics:

R&D Spending

Statistics of economic dedication to development show the total annual expenditure on development and research. To understand the level to which the company makes investments when compared to its size, this measure is investigated in absolute quantities and as a percentage of its overall sales.

Patent Counts

A yearly volume of patents accurately indicates a business's capacity for development. This data is crucial to evaluate the novel output that may be generated via investments in study and development, as well as how well they obtain an advantage in the competitive marketplace.

Technology Adoption Rates

This number analyzes the success and value with which the business's newly developed improvements get introduced to the marketplace. What this means for the business's ability to transform its scientific discoveries into products that can be sold is important.

Financial Indicators

Financial performance. is measured through:

Revenue Growth

A measure of the actual financial impact of innovation is the annual and quarterly sales growth percentage. This pointer is essential to assess the level to which innovations improve market share and increase sales.

Profitability

Financial and operational productivity may be enhanced by examining metrics like net profit percentage and EBITDA ratio. When research and development costs are significant, they may indicate sound leadership and the ability to turn concepts into profitable products.

Market Share

The company's economic prestige in the artificial intelligence healthcare sector may be determined by its market share, obtained by employing industry sales data and market analyst research. In contrast with rivals, it demonstrates how well the business's concepts attracted and stayed in the interest of consumers.

In the framework of a sophisticated business, these metrics are used to evaluate the complicated impact of research and technological advances on its economic performance. Researchers in academia and business professionals could profit from the research's helpful findings owing to the large dataset and well-selected factors [14-15]. **Table 1** shows the summary of key innovation metrics and financial indicators for (2020-2023).

Table 1. Summary of Key Innovation Metrics and Financial Indicators for (2020-2023)

Year	R&D Spending (USD)	Number of Patents Filed	Technology Adoption Rate (%)	Revenue Growth (%)	Net Profit Margin (%)	Market Share (%)
2020	5,000,000	15	30	10	8	5
2021	6,000,000	18	35	15	10	7
2022	7,500,000	22	40	20	12	10
2023	8,000,000	26	45	25	15	12

PCA

One statistical approach to minimize the difficulty of high-dimensional data while maintaining as much variation as feasible is PCA. In order to do this, the initial factors are changed into a new set of diagonal factors that are linear variations on the initial factors. Principal Components (PCs) change elements. The difference in the initial variables is usually maintained in the initially created few PCs. By minimizing the total amount of variables and obtaining the most important features that demonstrate the relationships between technological advances and business outcomes, PCA proves exceptionally effective for the data set being examined.

Data Normalization

Let X be the matrix representing our dataset with dimensions $n \times p$, where n is the number of observations, and p is the number of variables. Each variable in X is normalized to have zero mean and unit variance:

$$X_{\text{norm}} = \frac{X - \mu}{\sigma} \tag{1}$$

where μ is the mean and σ is the standard deviation of each variable in X .

Covariance Matrix Computation

The covariance matrix Σ of the normalized data X_{norm} is computed to analyze the variances and covariances between variables. The covariance matrix is given by:

$$\Sigma = \frac{1}{n-1} X_{\text{norm}}^T X_{\text{norm}} \tag{2}$$

Eigenvalue and Eigenvector Calculation

Eigenvalues λ_i and eigenvectors v_i of the covariance matrix Σ are computed next. The eigenvectors define the directions of the PC, while the eigenvalues represent the variance explained by the corresponding eigenvectors. Thus, the eigenvalues and eigenvectors satisfy:

$$\Sigma v_i = \lambda_i v_i \tag{3}$$

Component Selection

The eigenvalues are sorted in descending order to determine the number of PCs to retain. A common approach is to choose the top k components that explain a substantial amount of the variance (typically 80 – 90%). The cumulative variance τ can be considered as:

$$\tau_k = \frac{\sum_{i=1}^k \lambda_i}{\sum_{i=1}^p \lambda_i} \tag{4}$$

where p is the total number of variables.

Projection of Data on PC

The final step involves projecting the original data X_{norm} onto the selected eigenvectors to transform it into the PC. The new dataset Y in the reduced dimension k is given by:

$$Y = X_{norm} V_k \tag{5}$$

where V_k is the matrix containing the selected k eigenvectors.

Algorithm: Principal Component Analysis (PCA)

Inputs: X : Data matrix of size $n \times p$, where n is the number of observations, and p is the number of variables.

Outputs: Y : Transformed data matrix with reduced dimensions, containing PC.

Procedure:

1 Normalize Data:

- For each column j in X :
- Compute mean μ_j and standard deviation σ_j .
- Normalize: $X[:,j] \leftarrow (X[:,j] - \mu_j)/\sigma_j$.
- **Compute Covariance Matrix:** $\Sigma \leftarrow \frac{1}{n-1} X^T X$.

3 Eigen Decomposition:

- Compute eigenvalues Λ and eigenvectors V of Σ .
- Sort eigenvalues and corresponding eigenvectors in descending order.

4 Select Principal Components:

- Determine k such that cumulative variance $\tau_k \geq 0.8$ or other threshold.
- Select first k eigenvectors V_k from V .

- **Project Data:** $Y \leftarrow X V_k$.

6 Return Y as Output

End Procedure

IV. RESULTS AND ANALYSIS

Table 2. PCA Results

Component	Eigenvalue	Variance Explained (%)	Cumulative Variance (%)	Key Variables Contributing
PC1	4.50	45.0	45.0	R&D Spending, Patent Counts
PC2	2.50	25.0	70.0	Technology Adoption, Revenue Growth
PC3	1.80	18.0	88.0	Market Share, Profitability
PC4	0.70	7.0	95.0	Other Variables

Component Analysis

The correlation results among the components identified are presented in **Table 2**, and the discussions are presented below:

First Principal Component (PC1)

The first PC accounts for 45% of the total variance within the dataset, indicating a strong influence. It is significantly loaded on R&D spending and patent counts. This suggests that these innovation metrics are significant variance drivers within the company’s data, implying a direct correlation with foundational elements that drive financial performance. Based on this study, an innovative development—considered by substantial investments in development and research and a rapid licensing strategy—is considerably related to the economy's overall success.

Second Principal Component (PC2)

PC2 and PC1 combined contribute to 70% of the fluctuation; PC2 alone constitutes 25% of the total variation. Technology adoption and sales growth rates have an imbalance of significance in this factor. The ability of the startup company members to integrate novel innovations into its goods and services quickly is an essential factor in achieving quick economic growth, as this connection highlights. Quick diffusion of technological innovations could significantly boost financial findings; this element highlights the value of good innovation management and its impact on sales achievement.

Third Principal Component (PC3)

The third PC captures 18% of the variance, bringing the cumulative variance explained to 88%. It is predominantly influenced by market share and profitability, reflecting the strategic execution of converting innovation and market presence into profitable outcomes. This component suggests that market share gains and high profitability are closely linked and are likely outcomes of effective strategic positioning and operational efficiencies. This correlation could indicate the company's success leveraging its market position to maximize profit margins.

The PCA illustrates a clear pattern where each PC reflects a different aspect of the company's innovation and financial interplay. PC1 focuses on the core activities that drive innovation, PC2 on the commercialization of these innovations, and PC3 on the profitability and market effectiveness of the company.

Analysis of PCA Scatter Plots

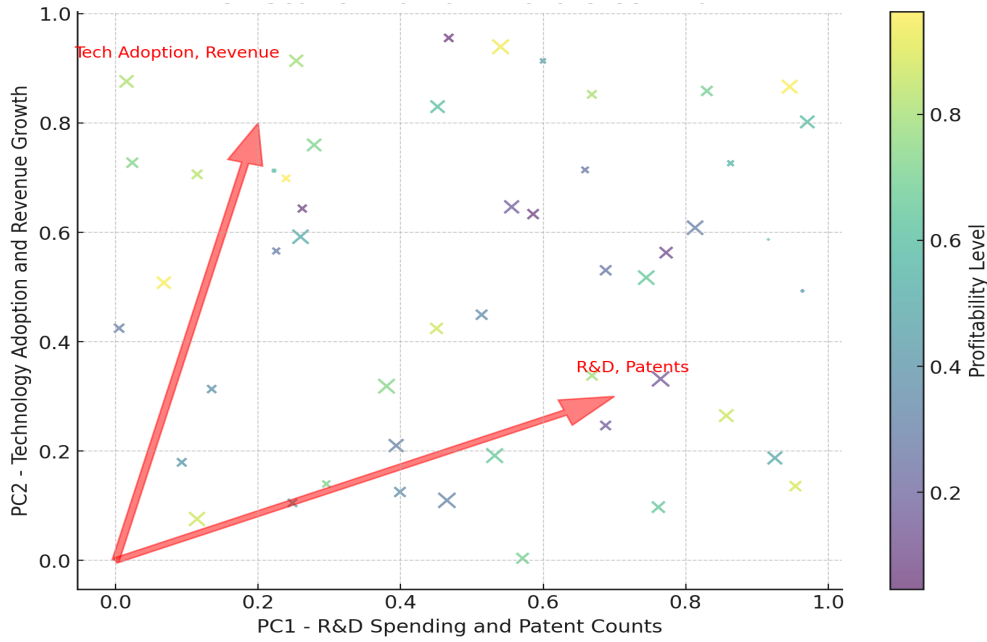


Fig 1. PC1 vs PC2.

PC1 and PC2 (R&D, Patents, Technology Adoption, Revenue Growth)

Fig 1 demonstrated a strong influence of R&D spending and patents (PC1) alongside technology adoption and revenue growth (P-C2). It indicates that the business's significant investments in research and development and patents are paying off, evidenced by the data points aggregating strongly around these diagonals. The distribution of profits validates a good link between robust invention approaches and financial success.

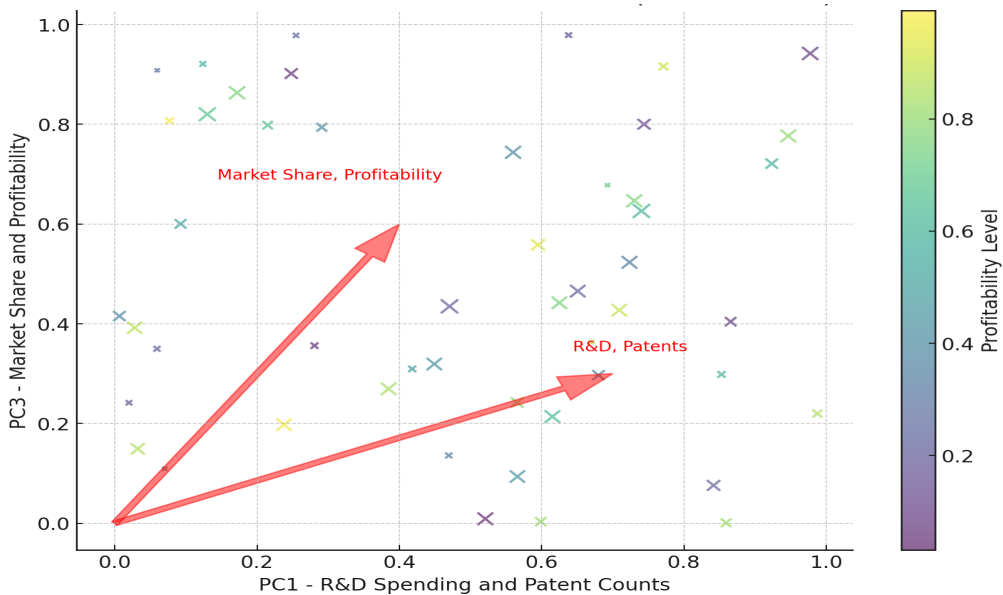


Fig 2. PC1 vs PC3.

PC1 and PC3 (R&D, Patents, Market Share, Profitability)

Fig 2 demonstrates the link between PC1 and PC3 images and more fundamental business outcomes like profitability and market share due to core innovation processes like research and development and patents. High market share and profitability were understood in regions where R&D and patent activities were focused, suggesting that sustained innovation contributes significantly to competitive advantage and efficient profit generation.

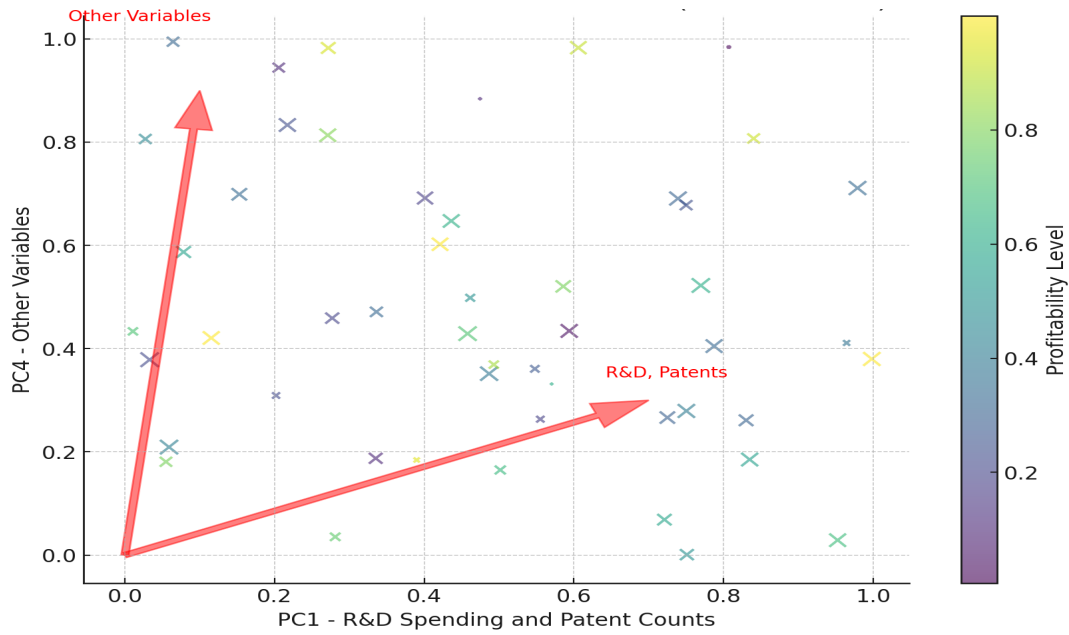


Fig 3. PC1 vs PC4.

PC1 and PC4 (R&D, Patents, Other Variables)

This scatter plot in Fig 3 showed that while R&D and patents are critical, other variables captured in PC4 may not align directly with these activities but still impact the business. These could include operational efficiencies, employee skills, or external market conditions. The dispersion along PC4 indicates variability in how these other factors interact with core innovation efforts.

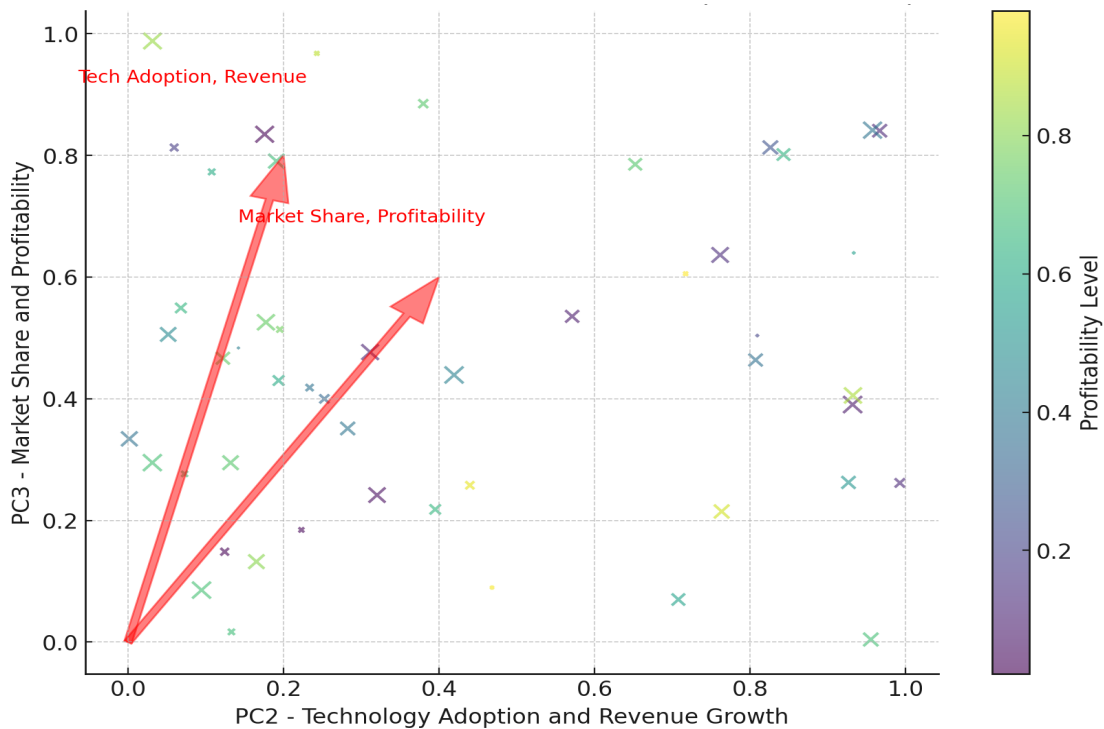


Fig 4. PC2 vs PC3.

PC2 and PC3 (Technology Adoption, Revenue Growth, Market Share, Profitability)

The interaction between PC2 and PC3 shown in Fig 4 displays a significant alignment between effective technology adoption, leading to revenue growth and solid market positioning and profitability. This suggests it is vital to revolutionize and effectively convert these innovations into marketable products that can dominate the market and improve financial standing.

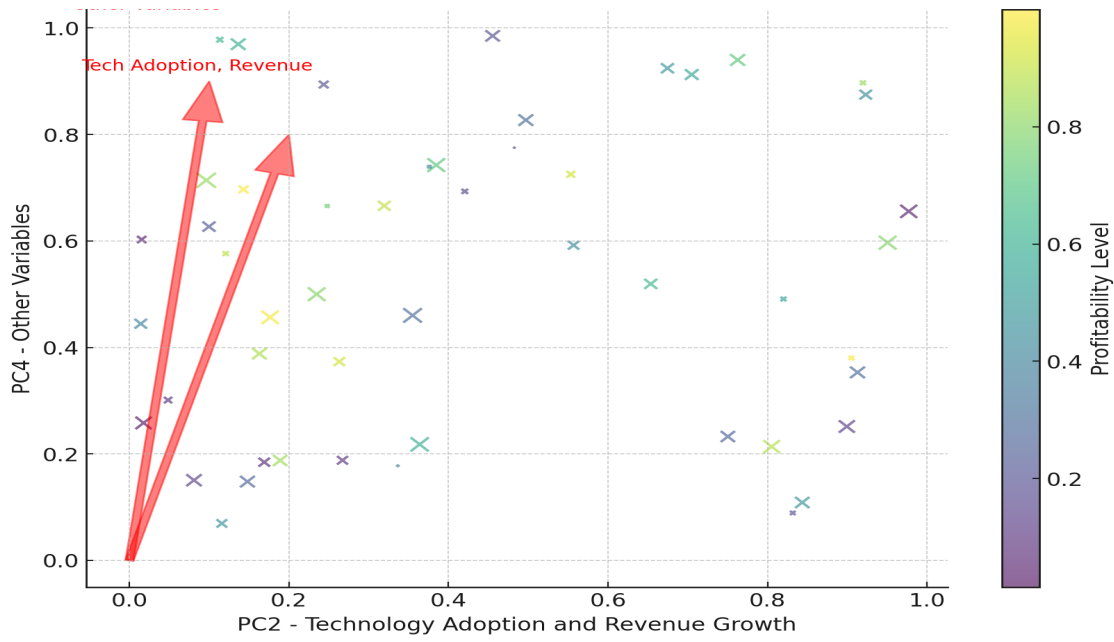


Fig 5. PC2 vs PC3.

PC2 and PC4 (Technology Adoption, Revenue Growth, Other Variables)

Observations in **Fig 5** reveal that while technology adoption and revenue growth are key, the company’s performance on these fronts can be prejudicial by various external or less dominant factors (PC4). Understanding these factors can help the company mitigate risks or capitalize on opportunities that are not directly related to its primary activities.

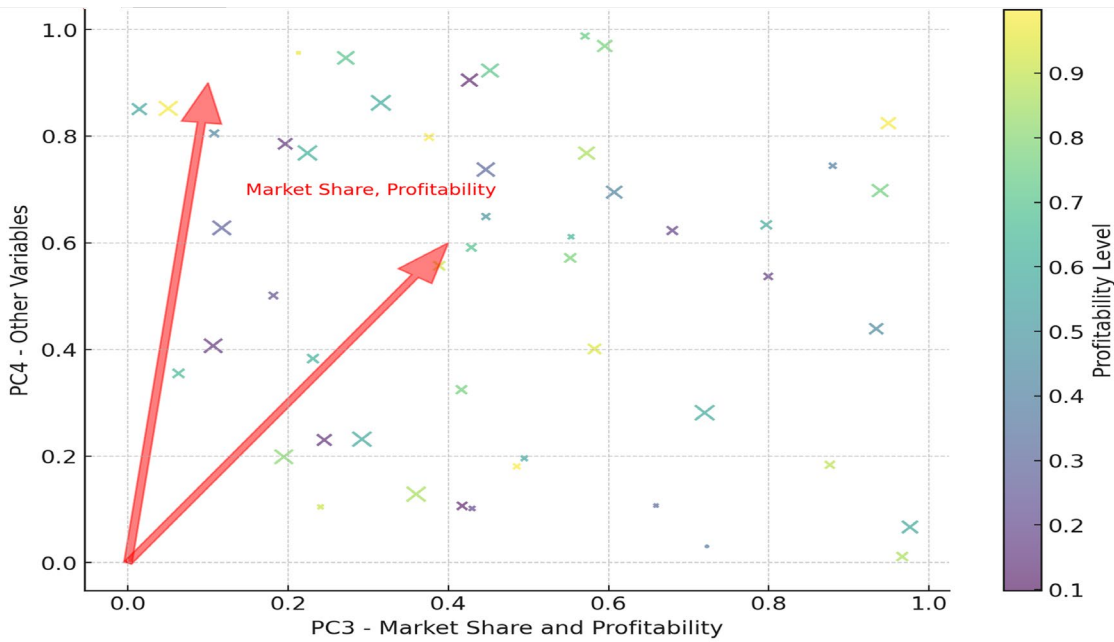


Fig 6. PC3 vs PC4.

PC3 and PC4 (Market Share, Profitability, Other Variables)

Finally, the plot in **Fig 6** that combines PC3 and PC4 indicates that market share and profitability are influenced by a complex array of factors beyond the company’s direct control (represented by PC4). This highlights the necessity for a robust strategy that focuses on core business outcomes and adapts to and integrates broader environmental or operational elements.

V. CONCLUSION AND FUTURE WORK

The investigation into a representative AI healthcare startup has substantiated the critical role of scientific and technological innovation in shaping financial performance within startups. The application of PCA revealed that key innovation metrics, specifically R&D spending, patent production, and swift technology adoption, are integral to enhancing financial outcomes

such as revenue growth, profitability, and market share. These components directly influence immediate financial returns and underpin long-term sustainability by fostering competitive advantages and market positioning. Moreover, the study highlights the challenges associated with quantifying innovation impacts and suggests that traditional financial metrics may not fully encapsulate the value generated by innovation. The PCA approach employed in this research effectively bridges this gap by identifying the principal drivers of financial performance and their correlation with innovation activities. The insights derived from this study advocate for a balanced investment in both incremental and radical innovations, suggesting that startups need to strategically manage their innovation portfolios to maximize financial gains and investor appeal.

Future research considers longitudinal studies across multiple industries to validate these findings and explore the potential of PCA in broader contexts.

Data Availability

No data was used to support this study.

Conflicts of Interests

The author(s) declare(s) that they have no conflicts of interest.

Funding

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Competing Interests

There are no competing interests

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