# Predicting Startup Funding Success- A Machine Learning Approach

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#### **Business Problem**

The challenge of obtaining funding is a significant barrier for emerging companies. A predictive model that accurately forecasts funding outcomes can significantly impact the efficiency of the investment process, benefiting both startups and investors by streamlining their approach to fundraising and investment, respectively.

# Background

The startup funding world has changed a lot. More money is out there, but it is also more challenging for startups to stand out and get the funding they need. This is because getting investment is a complex process. There is a real need for tools that can help predict whether a startup will succeed in getting funding. This would help both startups and investors make better decisions.

### **Data Explanation**

The project utilizes the "Startup Investments Crunchbase" dataset, encompassing data from approximately 54,000 startups, including details on funding rounds, investors, and industry classifications. Data preparation involved extensive cleaning and preprocessing to ensure quality and consistency for analysis.

- > Data Prep and Key Data Points
  - Funding Rounds: Number of successful funding rounds a startup has completed.
  - Venture Funding: Amount of venture capital received.

- **Startup Market**: The startup industry segment is categorized into 43 distinct groups.
- **Age at First Funding**: Time from inception to the first funding round, averaging 40.57 months.

# Methodology

The study employed logistic regression and decision tree models for their efficacy and interpretability in classification tasks. Model performance was evaluated based on accuracy, precision, recall, and F1 score, with cross-validation ensuring robustness.

### **Preliminary Analysis**

- Exploratory Data Analysis (EDA)
- An average of 40.57 months to first funding, with a funding range of 1-10 million dollars.
- Sectors such as software, biotechnology, and e-commerce dominate, with a significant concentration of startups in the United States.
- ➤ Model Performance
- Logistic Regression demonstrated superior performance with an accuracy of 83.58% and precision of 84.73%.
- Decision Tree Classifier showed an accuracy of 78.99%, with precision and F1 scores indicating strong predictive capability.

#### Conclusion

The startup funding scene has evolved significantly. While more investment is available, it is more challenging for new companies to grab attention and secure funds due to the complicated

nature of obtaining investments. There is a growing demand for tools that can forecast a startup's success in acquiring funding, aiding both startups and investors in making smarter choices.

# **Assumptions and Limitations**

The analysis relies on the dataset to represent the startup ecosystem accurately, emphasizing key features' impact on funding while acknowledging its limitations and the complexity of the funding process, which might omit certain influential factors.

### Challenges

Addressing missing data, managing outliers, and ensuring model accuracy without overfitting represented significant challenges. Moreover, selecting predictive features required a deep dive into the dataset, balancing statistical significance with practical relevance.

### **Future Directions and Recommendations**

Further research could explore additional machine learning techniques and integrate more comprehensive features, such as team dynamics and marketing strategies, to enhance predictive accuracy. The Logistic Regression model is recommended for stakeholders, combining high accuracy with interpretability for practical application.

### **Implementation Plan**

An ongoing validation strategy against new datasets is proposed, including iterative training to adapt to market trends, integration into investment tools, and the development of an accessible interface for evaluating funding prospects.

# **Ethical Considerations**

Ensuring data privacy, mitigating biases, and transparently communicating the predictive model's limitations are paramount. The goal is to foster an ecosystem where predictive analytics promote fairness, diversity, and innovation among startups.