Government Contracts in the Pandemic Era: A Comprehensive Impact Analysis Using Predictive Analytics



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ABSTRACT

In times of crisis such as the COVID-19 pandemic, effective allocation of resources becomes paramount worldwide. Particularly, the procurement of goods and services through government contracts plays a pivotal role in disaster management This capstone project, conducted at the University of Illinois Chicago, focuses on exploring government spending trends and predicting winning business types for government contracts amidst the COVID-19 pandemic. Predictive analytical models like KNN, Random Forest, Decision Tree, XG Boost, Linear Regression & LSTM were used as well as a UI has developed using Streamlit. By providing insights into government spending patterns successful contract duration and value, this research aims to significantly businesses' decision-making during crisis along with assisting them in gauging successful bids for government contracts, thereby facilitating informed decision-making and resource allocation strategies for effective disaster management.

BUSINESS PROBLEM

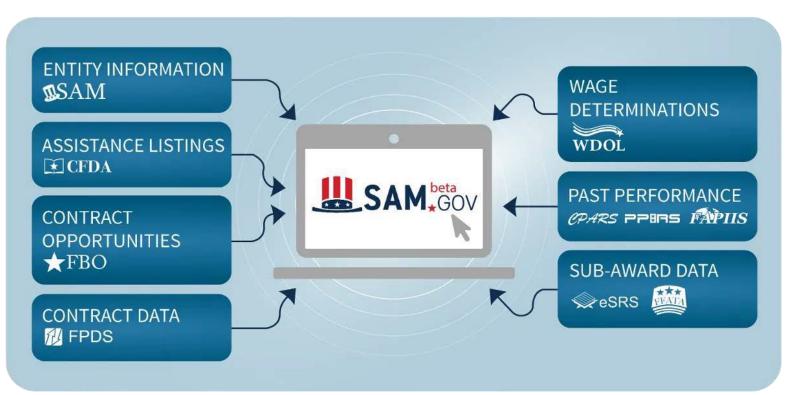
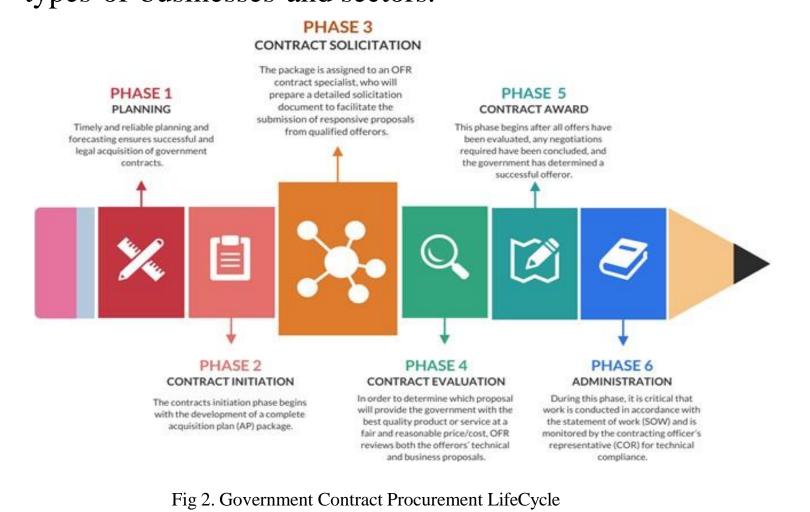


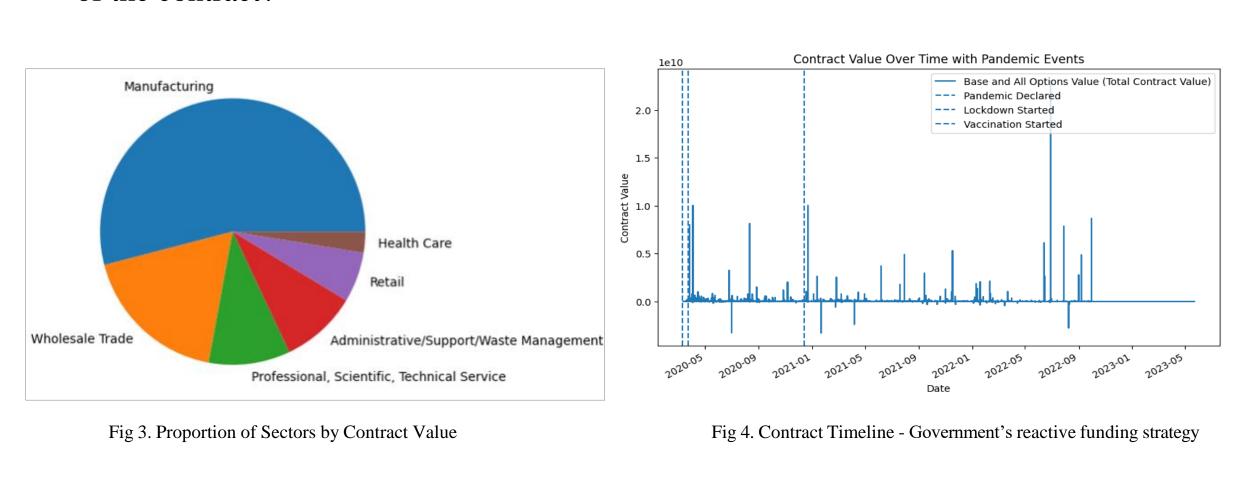
Fig 1. SAM.gov: US Government Contract Opportunities Database

contracts process typically including evaluation, and submission, pandemics or natural disasters, this process can be disrupted due to various factors such as budget reallocations, shifts in government priorities, and resource constraints. These disruptions may lead to delays in contract solicitations, cancellations of ongoing procurement processes, or changes in the types of goods and services needed by government agencies. The impact of such disruptions varies across different types of businesses and sectors.



RESEARCH QUESTIONS

- What factors drive the success of diverse business types (e.g., Women Owned, Small Business) in securing government contracts?
- How effective are machine learning algorithms in predicting winning business type for government contracts, particularly in multiclass classification scenarios?
- How effective are machine learning algorithms in predicting the duration and value of the contract?



ANALYTICAL PROBLEM

Model 1: Predict Business Type Winning the Contract Using different multiclass classification models like KNN, Random Forest, Decision Tree, we would predict the business type that can win the contract

Model 2: Predict Contract Duration Using Regression & XG Boost, we would predict the contract duration

Model 3: Predict Contract Value Using LSTM & Prophet, we would predict the contract value

DATA

The data for this study was sourced from sam.gov, comprising 150,000 government contracts with a total of 174 attributes. In the data cleaning phase, 110 relevant columns were selected, while contracts with negative or zero values were eliminated, along with any entries containing null values. To prepare the data for analysis, all text columns were label encoded, and 'Business Types' with less than 0.5% occurrence were dropped. Furthermore, highly correlated columns were removed to prevent multicollinearity. This rigorous data processing ensures that the dataset is suitably refined, optimized for subsequent analysis and model development, enhancing the accuracy and reliability of the research findings.

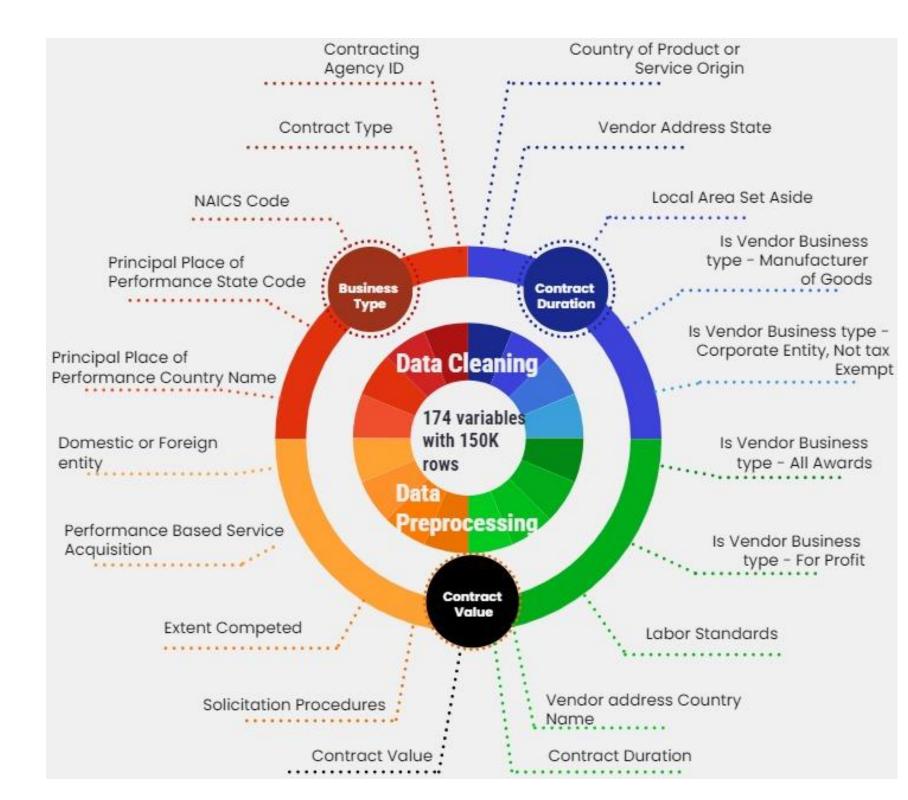


Fig 5. Input Variables

Fig 6. Technologies and Tools Used Model Selection Model Testing

Models used for the text columns lead Small predicting Business Business Type Selected 110 Dropped the Businesses Type - KNN, Random Forest Decision Tree and variables from 174 'Business Types thrive. Contract Duration which had less VA.MA and NY Random Forest Removed negative than 0.5% top in number Models used for Predicting and zero value occurrence in data of contracts predicting Contract Manufacturing **Duration-Linear** Contract Value Dropped Null and Health are Regression and the key sectors. XGBoost Models used for predicting Contract Value - LSTM and

*DoD: Department of Defense *HHS: Health and Human Services

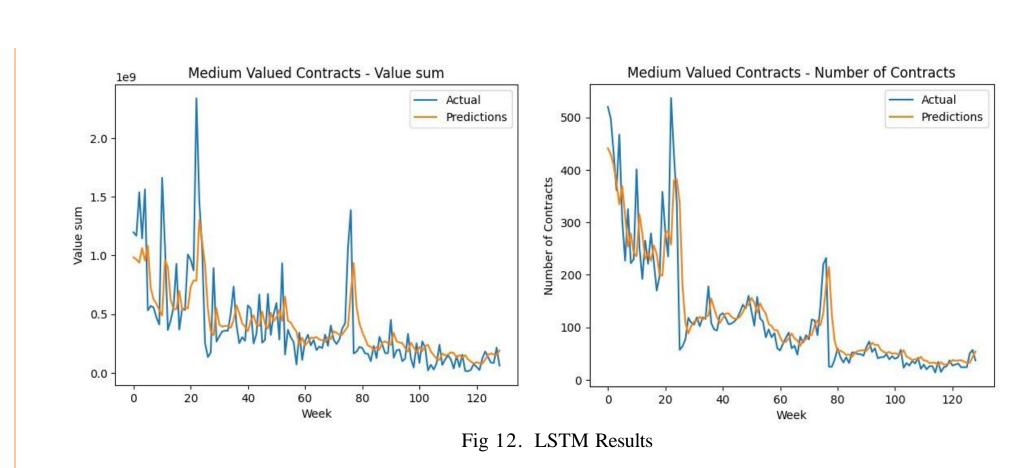
Fig 7. Methodology Diagram

MODEL BUILDING

METHODOLOGY

- For predicting business type utilized: Decision Trees, Random Forest, KNN; with Random Forest leading at 92.5% accuracy, proving highly effective in forecasting the successful business type winning government
- In predicting contract duration, XGBoost model demonstrated superior performance with R² value of 0.717, indicating a robust fit, while Linear Regression showing limited predictive power with R^2 of 0.207.
- Employed Long Short-Term Memory (LSTM) networks to accurately predict weekly contract counts and values, leveraging sequence-based analysis of 5 or 8-week periods. Categorizing contracts into three value ranges prior to the analysis was needed to enhance predictive efficacy.





DEPLOYMENT & LIFE CYCLE MANAGEMENT

- Documentation Detailed documentation with the business problem and the solution used to solve it have been provided along with a paper highlighting the project life cycle.
- Source Code GitHub repository has been provided with all the necessary code developed to solve the business problem.
- User Interface Created a User Interface using Streamlit to take contract inputs for providing predictions. Tableau visualizations have been created and integrated with the UI.

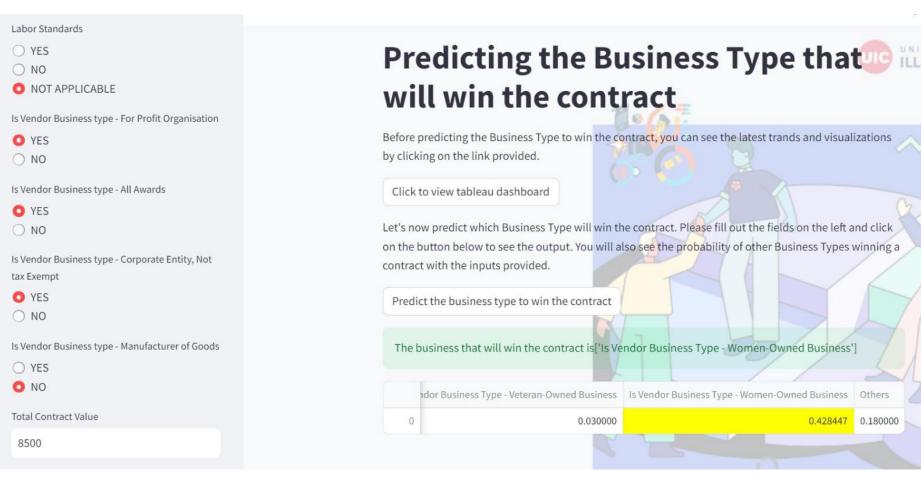


Fig 13. User Interface for Business Usecase

CONCLUSION

Our aim was to help vendors understand the nature of government contracts during a pandemic. To achieve this, we have created visualizations to help gain insights, like places which are giving out most contracts, vendor locations that are most successful in getting the contracts and so on. The models built, help predict the business type most likely to win the contract along with the most likely contract value and duration. This would help the vendors be better prepared for the upcoming contracts.

We would further like to work and make this a one stop shop for vendors by adding more features like predicting the number of contracts released in a particular sector and adding features like connecting multiple vendors to complete a contract.

ACKNOWLEDGEMENT

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