

Canada Revenue Agency Project Cost Difference & Delivery Success

*

Government of Canada

Gouvernement du Canada

Canada Revenue Agency

Forecast

Ang Li, Bo Yang, Li Zhuang, Xiaomeng Xu, Ziang Cui Business Intelligence System Infrastructure, Algonquin College

Introduction

In today's dynamic market, accurate budget forecasting and efficient cost control are essential for project success. Our client, CRA, seeks a comprehensive analysis of projected budget estimates to align financial planning with actual expenditures. Thus, our project aims to utilize advanced machine learning and Power BI analytics to achieve three key objectives: predicting disparities between estimated and actual costs, forecasting project delivery success probabilities, and generating data-driven insights for stakeholders. These objectives aim to enhance financial accuracy, improve project outcomes, and provide decision-makers with actionable intelligence for a competitive edge in project management.

Tools

• JIRA Project Management

• Power BI Dashboard Creation

Python Machine Learning

Data

12 Columns * 64 Rows Synthetic Data

Process

Power BI

- Analysis Direction: Determine focus areas based on results and data.
- Data Examination: Use Power BI for detailed data analysis.
- **Dashboard Creation**: Compile insights into an informative dashboard.

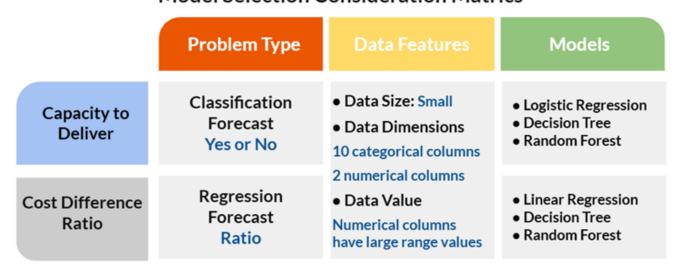
Machine Learning

- **Prediction Type**: Classify prediction results as classification, regression, or both.
- Model Research: Investigate models, considering strengths and limitations.
- Data Cleaning: Remove inaccuracies, handle missing data, and normalize data.
- Initial Predictions: Use selected model for initial predictions.
- Parameter Tuning: Adjust model parameters for optimization.
- **Testing Adjustments**: Validate model adjustments with additional predictions.

Data	Observ	ation Inp	ut/Outp	ut Configu	ıration	Model Tuning	C	onclusion
		Model Sele	ection		Model Build Up		Model Eval	uation
	0%	10%	20%	30%	60	% 80	% 90%	100%

Machine Learning Models

Model Selection Consideration Matrics



Result Power BI



 High-priority projects show greater success rates, underlining the importance of strategic prioritization. Variations in departmental performance suggest differing management practices and resource allocations.

	Priority	Work Classification	Department
Major influencer	▽	×	▽
Impact multiplier	6.19	-	41.25

X Capacity of Deliver could be Predict well

 Cost variances are not significantly influenced by individual managers, application names, or task numbers, indicating a complex interplay of factors contributing to budget deviations within projects.

	Manager	Application Name
Major influencer	×	×
Impact multiplier	-	-

X Cost Difference could NOT be Predict well

Machine Learning

Output=capacity of deliver							
	Logistic Regression	Decision tree	Random forest model				
Accuracy	0.95	0.84	0.94				
Precision	0.98	0.94	0.97				
Recall	0.96	0.87	0.97				
f1 score	0.97	0.92	0.96				
Best Model	√						
Output=cost difference ratio							
	Linear Regression	Decision tree	Random forest				
MSE	8.01	7.74	8.10				
MAE	1.59	1.08	1.20				
R2	-3.74	-0.41	-0.46				
Best Model		√					

Conclusion

Capacity to Deliver:

- Power BI analysis reveals that high-priority projects exhibit higher success rates, emphasizing the importance of strategic prioritization.
- Machine learning models demonstrate the robustness of the generalized model for Capacity to Deliver, indicating its reliability in determining project success across various scenarios.

Cost Difference Ratio:

- Power BI analysis suggests that individual factors such as managers, application names, or task numbers have limited influence on cost variances.
- Machine learning findings highlight the challenge of generalizing models for Cost Difference Ratio due to the complexity of relationships between inputs and the limited volume of data available.

Recommendations

Improve Data Quality: Establish standardized procedures for data collection, recording, and maintenance to enhance data quality and consistency.

Cautions

Exercise caution with datasets of uncertain quality and small sizes, as they can significantly affect model reliability and effectiveness.

Lessons to Learn

- Re-evaluate Modeling: Assess and adjust existing models or explore alternatives for larger datasets. Refine model parameters for improved accuracy.
- **Continuous Improvement**: Embrace ongoing improvement, incorporating larger datasets and refining modeling strategies while learning from past challenges.