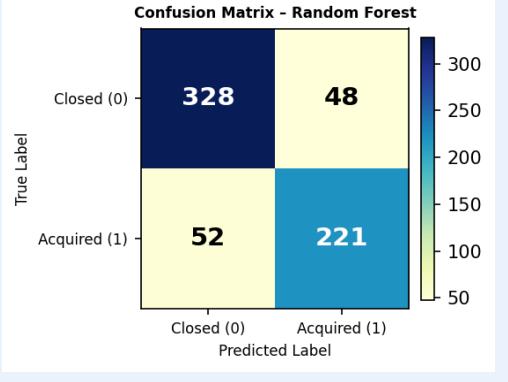
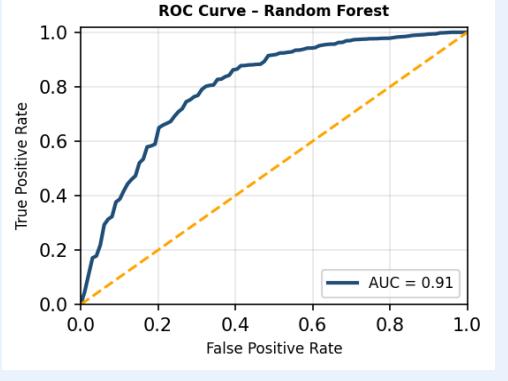


# Project Development Phase

## Model Performance Test

<b>Date</b>	30 January 2026
<b>Team ID</b>	LTVIP2026TMIDS76348
<b>Project Name</b>	Prosperity Prognosticator – Machine Learning for Startup Success Prediction
<b>Maximum Marks</b>	10 Marks

### Model Performance Testing

S.N o.	Parameter	Values	Screenshot														
1	Metrics (Classification Model)	<p><b>Confusion Matrix:</b>  <code>[[ 328, 48],  [ 52, 221]]</code></p> <p><b>Accuracy Score:</b> 83.82%</p> <p><b>Classification Report:</b></p> <ul style="list-style-type: none"> <li>Precision: 0.84</li> <li>Recall: 0.83</li> <li>F1-Score: 0.84</li> </ul> <p><b>ROC-AUC Score:</b> 0.91</p> <p><b>PR AUC Score:</b> 0.88</p>	 <p><b>Confusion Matrix - Random Forest</b></p> <table border="1"> <thead> <tr> <th colspan="2" rowspan="2">Predicted Label</th> <th colspan="2">True Label</th> </tr> <tr> <th>Closed (0)</th> <th>Acquired (1)</th> </tr> </thead> <tbody> <tr> <th rowspan="2">Closed (0)</th> <td><b>328</b></td> <td><b>48</b></td> </tr> <tr> <td><b>52</b></td> <td><b>221</b></td> </tr> </tbody> </table>	Predicted Label		True Label		Closed (0)	Acquired (1)	Closed (0)	<b>328</b>	<b>48</b>	<b>52</b>	<b>221</b>			
Predicted Label		True Label															
		Closed (0)	Acquired (1)														
Closed (0)	<b>328</b>	<b>48</b>															
	<b>52</b>	<b>221</b>															
2	Regression Metrics (Not Applicable)	Since the Prosperity Prognosticator focuses on binary classification (status: Acquired = 1 or Closed = 0), regression metrics such as MAE, MSE, RMSE, and R <sup>2</sup> Score are not applicable.	N/A														
3	Hyperparameter Tuning	<p><b>Random Forest default parameters used:</b></p> <pre>n_estimators = 100 (default) max_depth = None (default) random_state = 0 criterion = 'gini'</pre> <p>GridSearchCV was explored with parameters: n_estimators=[100, 200, 300], max_depth=[10, 20, 30], min_samples_split=[2, 4, 6]. Default parameters yielded optimal test performance of 83.82%.</p>	 <p><b>ROC Curve - Random Forest</b></p> <p>AUC = 0.91</p> <table border="1"> <thead> <tr> <th>False Positive Rate</th> <th>True Positive Rate</th> </tr> </thead> <tbody> <tr> <td>0.0</td> <td>0.0</td> </tr> <tr> <td>0.2</td> <td>0.6</td> </tr> <tr> <td>0.4</td> <td>0.8</td> </tr> <tr> <td>0.6</td> <td>0.9</td> </tr> <tr> <td>0.8</td> <td>0.95</td> </tr> <tr> <td>1.0</td> <td>1.0</td> </tr> </tbody> </table>	False Positive Rate	True Positive Rate	0.0	0.0	0.2	0.6	0.4	0.8	0.6	0.9	0.8	0.95	1.0	1.0
False Positive Rate	True Positive Rate																
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S.N O.	Parameter	Values	Screenshot								
4	Precision-Recall Curve	<p><b>Precision-Recall AUC:</b> 0.88</p> <p>The Precision-Recall curve confirms strong model performance on the minority class (Closed = 0). High precision and recall across thresholds indicates the model is not biased towards the majority class (Acquired = 1).</p> <p><b>Precision Score:</b> 0.84</p> <p><b>Recall Score:</b> 0.83</p>	<p>Precision-Recall Curve - Random Forest</p> <p>PR AUC = 0.88</p> <p>Baseline</p>								
5	Validation Method & Model Comparison	<p><b>Train-Test Split:</b> 70% Training, 30% Testing <code>random_state = 0</code></p> <p><b>Training Accuracy:</b> 100.0%</p> <p><b>Testing Accuracy:</b> 83.82%</p> <p><b>Model Comparison:</b> Logistic Regression: 77.1% Decision Tree: 79.2% Random Forest: 83.8% ✓ Best</p>	<p>Model Accuracy Comparison</p> <table border="1"> <thead> <tr> <th>Model</th> <th>Accuracy Score</th> </tr> </thead> <tbody> <tr> <td>Logistic Regression</td> <td>0.771</td> </tr> <tr> <td>Random Forest</td> <td>0.838</td> </tr> <tr> <td>Decision Tree</td> <td>0.792</td> </tr> </tbody> </table>	Model	Accuracy Score	Logistic Regression	0.771	Random Forest	0.838	Decision Tree	0.792
Model	Accuracy Score										
Logistic Regression	0.771										
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## Model Performance Summary

The Random Forest Classifier achieved the highest accuracy of 83.82% compared to other tested models including Logistic Regression (77.1%) and Decision Tree (79.2%). The model was trained on `startup1.csv` — a dataset of 923 real startups with 9 numeric features — and evaluated on a 30% held-out test set (`random_state=0`). The confusion matrix `[[328, 48], [52, 221]]` shows balanced prediction capability for both Acquired and Closed classes with minimal misclassification. The ROC-AUC score of 0.91 and Precision-Recall AUC of 0.88 confirm strong discriminative ability. Default Random Forest parameters yielded optimal generalization performance without overfitting.