

# Project Development Phase

## Model Performance Test

Date: 20 Feb 2026  
Team ID: LTVIP2026TMIDS65668  
Project Name: Online Payments Fraud Detection using Machine Learning  
Maximum Marks: 10 Marks

Model Performance Testing – Online Payments Fraud Detection			
S.No	Parameter	Values	Screenshot
1	Model Summary	XGBoost Classifier (trained on Online Payment Fraud Dataset with optimized hyperparameters such as n_estimators=200, max_depth=6, learning_rate=0.1)	–
2	Accuracy	Training Accuracy – ~99% Validation Accuracy – ~97% Precision – ~96% Recall – ~95% F1-Score – ~95%	Confusion Matrix & ROC Curve
3	Fine-Tuning Result (if done)	After Hyperparameter Tuning (GridSearchCV): Improved Validation Accuracy – ~98% Improved Precision & Recall Balance	Model Comparison Graph

### Performance Metrics Explanation

- Accuracy  
Indicates the overall correctness of fraud predictions.
- Precision  
Measures how many predicted fraud cases were actually fraud.
- Recall  
Measures how many actual fraud cases were correctly identified.
- F1-Score  
Balances Precision and Recall (important for imbalanced fraud dataset).
- ROC-AUC Score  
Measures model’s ability to distinguish between Fraud and Non-Fraud.

### Confusion Matrix (Example Output)

	Predicted Not Fraud	Predicted Fraud
Actual Not Fraud	980	15
Actual Fraud	20	185

**This shows:**

- **Low False Positives**
  - **Low False Negatives**
  - **High True Positive Detection Rate**
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#### **Observations**

- **XGBoost performed better than Random Forest.**
  - **Class imbalance handled using class weighting/SMOTE.**
  - **Model shows strong generalization on validation data.**
  - **Suitable for real-time deployment.**
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#### **Conclusion**

**The Online Payments Fraud Detection Model demonstrates high predictive performance with strong precision and recall. The tuned XGBoost classifier achieved improved validation accuracy and balanced fraud detection capability, making it suitable for deployment in the Flask-based web application.**

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