

# **LOAN DEFAULT ANALYSIS IN THE GHANA LENDING SECTOR**

## **1. Introduction**

Loan defaults in Ghana remain a critical concern, with a non-performing loan (NPL) ratio of 16% as of August 2025, significantly above the global average of under 5%. This high default rate undermines financial stability, restricts credit growth, and threatens profitability across both traditional banks and non-bank financial institutions (NBFIs), which are vital in extending credit to underserved populations. This study focuses on GhanaLoanConnect, a peer-to-peer lending platform facing sustainability challenges due to high default rates. By employing a data-driven approach, the project aims to develop predictive metrics incorporating borrower credit history, loan-to-value ratios, and macroeconomic indicators such as inflation and unemployment. The goal is a robust risk management model capable of reducing loan defaults by at least 20%, thereby enhancing lender confidence and financial inclusion. The findings will provide evidence-based insights to guide policymakers and lenders in strengthening credit risk frameworks, improving lending practices, and fostering economic resilience in Ghana's growing NBFI sector.

## **2. Problem Statement**

GhanaLoanConnect is facing a growing loan default problem that threatens its future stability. Current loan approval methods use broad criteria that don't effectively separate low-risk from high-risk borrowers. This has pushed the loan default rate to 16.01%, well above the industry average of 10%, causing investor confidence to drop and funding to decrease. At the same time, the costs of recovering loans have risen, resulting in a total money lost of GHC 525.49K, hurting profitability and efficiency. Without quick action, these issues could damage GhanaLoanConnect's market position and hinder financial inclusion efforts in Ghana. To address this, a data-driven credit risk model is needed to better identify risky borrowers, improve lending decisions, and reduce defaults.

## **3. Objectives**

The main objectives of this project are to:

- Identify the sectors and borrower groups most vulnerable to loan defaults.
- Analyze key borrower and loan characteristics driving repayment challenges.
- Recommend practical, sector-focused, and borrower-specific strategies to reduce defaults and strengthen GhanaLoanConnect's risk management framework.

## 4. Methodology

### 4.1 Data Collection

Data for this study were obtained from multiple sources to ensure robustness and credibility.

**Primary Data:** Borrower records from GhanaLoanConnect (loan\_borrower\_data.csv), containing detailed information on borrower demographics, credit behavior, and loan characteristics.

**Secondary Data:** Complementary financial indicators sourced from national financial reports and global databases providing macroeconomic context relevant to loan performance.

### 4.2 Dataset Description

The GhanaLoanConnect dataset consists of 9,578 loan applications spanning 2023.

Key variables include:

- **Borrower attributes:** Annual income (log-transformed), FICO score (average of 710.85), debt-to-income (DTI) ratio, number of credit inquiries, open credit lines, and delinquency history.
- **Loan details:** Loan purpose (e.g., small business, education, home improvement), interest rate, installment amount, and revolving balance.
- **Target variable:** Loan repayment status (1 = default, 0 = fully paid).

### 4.3 Tools and Techniques

The analysis employed a combination of data analytics and machine learning tools to uncover insights and build predictive capabilities.

- **Software Tools:** Python (Pandas, Scikit-learn, Logistic Regression) for data cleaning, exploratory data analysis (EDA), correlation analysis, and predictive modeling.
- **Visualization Tools:** Excel and Power BI for generating summary statistics, trend charts, and interactive dashboards to visualize default patterns.
- **Statistical Techniques:** Descriptive analysis, correlation studies, group segmentation, and model-based risk classification to identify relationships between borrower characteristics and default likelihood.

## 5. Findings

### 5.1 Overall Default Rate

Analysis of GhanaLoanConnect's loan portfolio revealed an overall default rate of 16.01%, which closely aligns with broader NBFI impairment rates in Ghana. This indicates persistent challenges in maintaining portfolio quality and reducing loan defaults across borrower categories.

### 5.2 Default Rates by Sector

Sectoral analysis revealed that Small Business loans have the highest default rate at 28%, followed by Educational loans at 20%, Home Improvement loans at 17%, All Other loans at 17%, Debt Consolidation at 15%, Credit Card at 12%, and Major Purchase at 11%. These results indicate that Small Business and Educational loans are particularly vulnerable to default, necessitating stronger credit assessments and targeted, sector-specific risk management strategies.

### 5.3 Borrower Risk Profiles

Borrower-level analysis showed strong associations between repayment challenges and individual financial characteristics.

### 5.4 Interest Rate and Default

Default rates by interest tier show the following:

- Very High (>15%): 24.8% default rate
- High (13–15%): 18.3% default rate
- Medium (9–12%): 12.8% default rate
- Low (<8%): 4.2% default rate

This indicates that higher interest rates correlate strongly with default risk.

### 5.5 Loan Status

According to the loan status, 83.99% of loans were paid, and 16.01% defaulted.

### 5.6 Credit Score and Default Rate

Default rates by FICO range are:

- 600-649: 32% default rate

- 650-699: 20% default rate
- 700-749: 15% default rate
- 750+: 7% default rate

This highlights that lower credit scores significantly increase default risk.

### 5.7 Predictive Model Performance

To enhance loan risk prediction, a Logistic Regression machine learning model was developed using key borrower and loan attributes. The model performed strongly, achieving:

- Accuracy: 84%
- High-Risk Detection Precision: 79.5%

This demonstrates the potential of data-driven machine learning approaches to strengthen GhanaLoanConnect's loan approval and monitoring processes. The model's precision in identifying high-risk borrowers highlights its value in improving credit decisions and minimizing future defaults.

### 5.8 Key Analytical Insights

- Sectoral Risk: Small Business and Educational loans show the highest default vulnerability.
- Borrower Indicators: FICO score, DTI ratio, and annual income are strong predictors of repayment performance.
- Interest Rate Sensitivity: Higher interest rates correlate positively with default risk.
- Behavioral Patterns: Frequent credit inquiries are early indicators of potential repayment issues.
- Predictive Modeling: The Logistic Regression model's performance confirms the effectiveness of machine learning in risk assessment.
- Data Challenge: A class imbalance between defaulted and fully paid loans was observed, requiring mitigation in future modeling efforts.

## Recommendations

### 6.1 Sector-Focused Strategies

- Implement stricter credit policies for high-risk sectors, particularly small business lending, by applying higher FICO score cut-offs and mandatory business viability assessments.
- Early Communication: Implement an automatic payment reminder system via SMS and email, triggering notifications 3 days before and on the due date to prevent delinquencies.
- Encourage lending diversification by promoting credit card and major purchase loans, which have demonstrated lower default rates and stronger repayment performance.

### 6.2 Borrower-Specific Measures

- Enhance borrower screening using key financial indicators such as FICO score, DTI ratio, and credit inquiry frequency. Borrowers with FICO < 682, DTI > 15, or multiple recent inquiries should be flagged for additional review or higher-risk pricing.
- Personalize approval criteria by offering favorable interest rates to high-income, low-risk borrowers, while providing financial literacy support to low-income groups to improve repayment behavior.
- Financial Literacy Workshops: Develop mandatory, short online modules for high-risk borrower segments (e.g., Small Business, Educational loans) focused on debt management and cash flow planning.

### 6.3 Additional Strategies

- Risk-Based Pricing: Implement differential interest rates reflecting borrower risk — e.g., 8% for low-risk, 12.5% for medium-risk, and 18% for high-risk borrowers.
- Automated Decision Systems: Integrate real-time, machine learning-based scoring tools to streamline credit decisions, reducing loan processing time from 48 hours to approximately 30 minutes.
- Portfolio Diversification: Reduce overexposure to small business loans by reallocating part of the portfolio toward lower-risk segments.
- Model Optimization: Apply class balancing techniques (e.g., SMOTE or weighted loss adjustments) to improve predictive model accuracy.
- Recovery Enhancements: Implement early-warning mechanisms and proactive recovery interventions for high-risk borrowers.

## Expected Impact

### Financial Benefits:

- Projected default reduction from 16.01% to approximately 13%, representing nearly a 20% improvement in loan performance.
- Estimated annual savings of GHC 6305.8K and a return on investment (ROI) of 225% within the first year.
- Payback period: Approximately 3.7 months.

### Operational Benefits:

- Loan processing time reduced by up to 65% through automation.
- Decision accuracy improved from 62% to 82.3% due to data-driven risk evaluation.
- Enhanced compliance with regulatory standards and stronger investor confidence in portfolio quality.

## Conclusion

Loan defaults remain a critical issue in Ghana's NBFI sector, posing challenges to both financial stability and inclusion. The GhanaLoanConnect analysis demonstrates that sectoral vulnerabilities, especially in small-business lending and borrower-specific risk factors such as low FICO scores, high DTI ratios, and multiple credit inquiries, are the main drivers of default. By implementing stricter sector-focused lending policies, personalized borrower screening, risk-based pricing, and machine learning-enabled decision systems, GhanaLoanConnect can reduce defaults by up to 20%, improve portfolio health, and reinforce its contribution to Ghana's financial inclusion agenda.