

# **White Box Finance: Interpreting AI Decisions in Finance through Rules and Language Models**

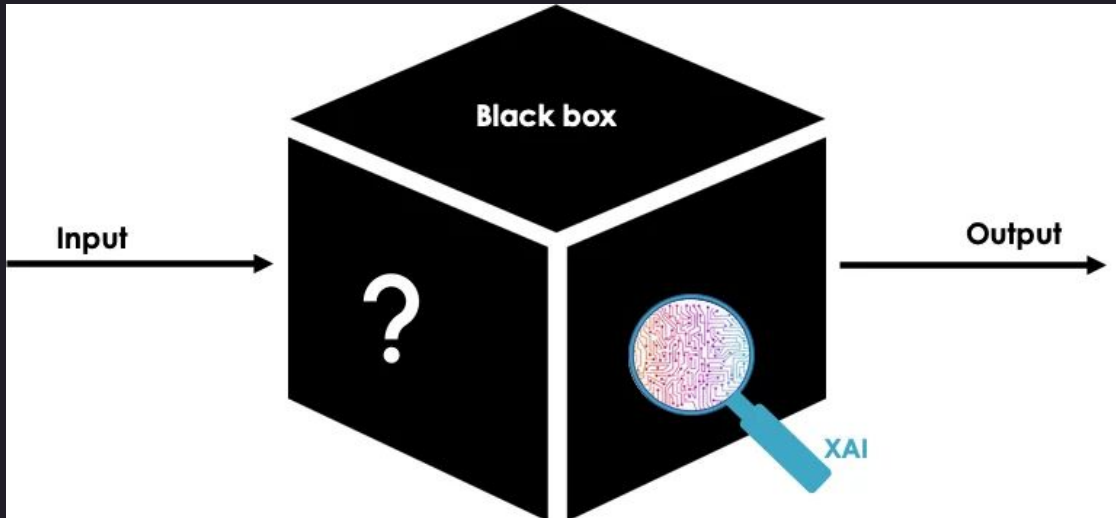
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# Motivation

Loan defaults → major financial losses.

ML models (e.g., XGBoost) improve prediction, but are black-boxes.

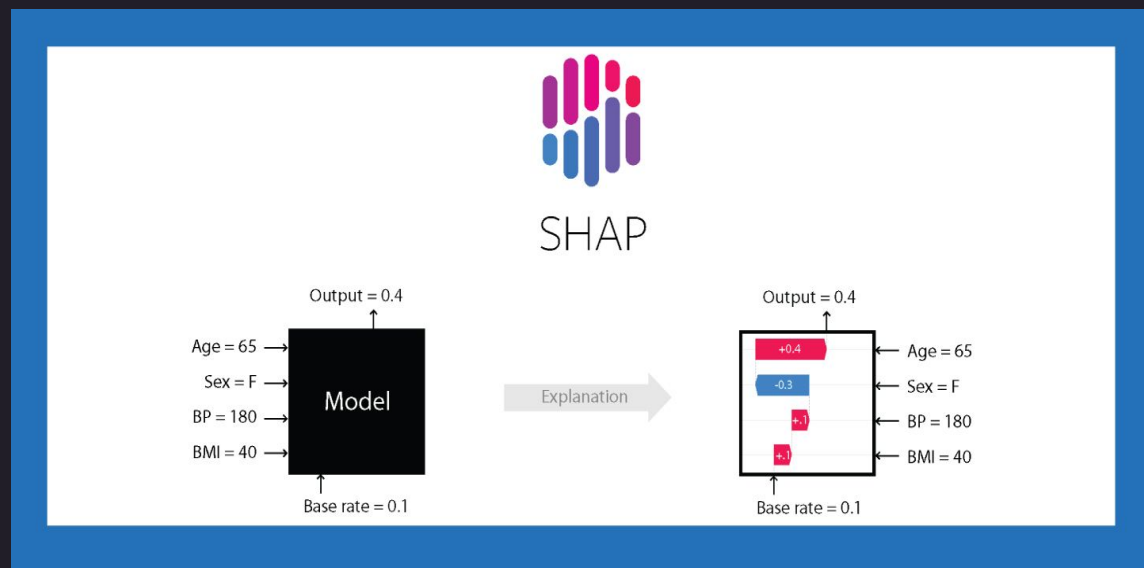
Finance requires transparent, auditable explanations for regulators, loan officers, and customers.



# Methodology

Enhance interpretability and trust in AI credit risk models by creating and comparing:

SHAP + GPT-4 → feature-based + natural language explanations.



# Methodology

Enhance interpretability and trust in AI credit risk models by creating and comparing:

Rule-based logic → transparent, business-aligned decision rules.



## Experimental Setup

- **Dataset:** Anonymized loan applicant records containing demographics, employment, credit history, and repayment behavior.
- **Preprocessing:** Missing values removed ( $<1\%$ ), categorical variables frequency-encoded, numerical features preserved.
- **Model:** XGBoost classifier trained with 5-fold stratified cross-validation. Class imbalance addressed using `scale_pos_weight`.
- **Evaluation Metrics:** Area Under the Curve (AUC), Precision, Recall, F1-score, and Confusion Matrix analysis

# Experimental Setup

## Explanation Modules

Two complementary explanation pipelines were applied to model predictions:

- **SHAP + GPT-4:** Local feature attributions → top 3–5 contributors → converted into business-friendly textual narratives.
- **Rule-Based Logic:** Categorical histograms and KDE plots used to derive interpretable decision rules aligned with institutional underwriting heuristics.

# Results

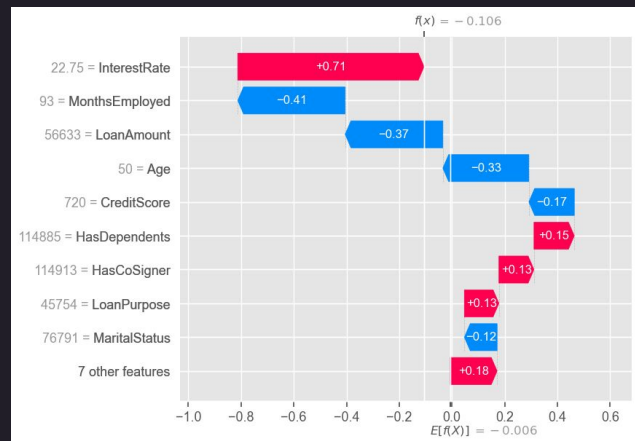
## Model Prediction

Age	Income	Loan Amount	Credit Score	Months Employed	Interest Rate	DTI Ratio	Education
36	80846	179949	347	20	23.96	0.9	PhD

# Results

## SHAP + GPT-4

The interest rate on the loan is quite high at 22.75%. This significantly increases the cost of borrowing, making it more challenging for the customer to manage their monthly payments. The high SHAP impact of 0.71 indicates that this factor is a strong contributor to the default risk



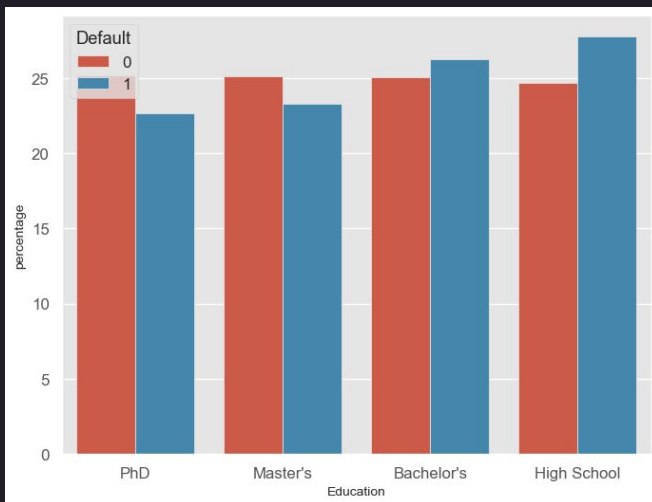


# Results

## Rule-Based Logic

if row["Age"] < 40:

    explanations.append("Young age may indicate lack of financial experience.")



## Conclusion and QA

- GPT Explanations → Rich, nuanced, human-friendly
- Rule-Based Explanations → Transparent, audit-ready, regulatory aligned
- Hybrid Approach = Best of both worlds: trust + compliance