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Predicting Loans with Machine Learning and Streamlit

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THE PROBLEM

Industry Overview



Most of the loan application process is laborious. The outcome is a simple 'Yes' or 'No' without further explanation.



Rejection is tough - don't let it make your business tougher. There is communication gap to be bridged.

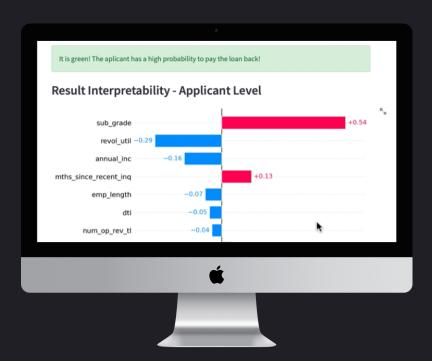
SOLUTION

Unlocking the Insights of Machine Learning

Despite its powerful capabilities, machine learning's charm was often hidden beneath technical jargon and the mystery of its "black box" operations.

The web application created with Streamlit used visual exploration to demonstrate the power of machine learning, showcasing the dynamic input feature and its SHAP value that drove the decision.

This improved the application's transparency and helped to close the communication gap.



What are SHAP Values?

- Calculation that represents the relative influence of each input features within the algorithm.
- It can be thought of as the 'weighting' that each feature had upon the final result of creditworthiness that the model predicted.

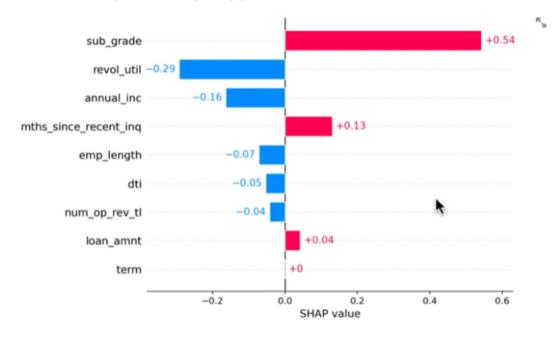
To predict default/ failure to pay back status, you need to follow the steps below:

- Enter/choose the parameters that best descibe your applicant on the left side bar;
- Press the "Predict" button and wait for the result.

Below you could find prediction result:

It is green! The aplicant has a high probability to pay the loan back!

Result Interpretability - Applicant Level



Model Interpretability - Overall

Why Should I Care About SHAP Values?

Unveiling the Creditworthiness Prediction:



SHAP values provide an intuitive way to understand the impact of individual features on a model's predictions



SHAP values can be used to identify important features and detect potential bias in a model



SHAP values can be used to improve model performance by identifying redundant or irrelevant features

MISSION

Helping customer to learn from their data



Lending Club dataset is used for model training

Available from Kaggle

Total observations

2.26m

Total features

150

Annual income:

65k

Loan amount (median):

\$13k

Avg interest rate

13%

Selected features

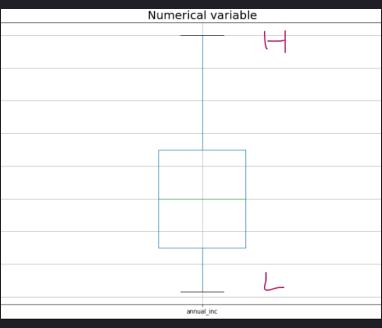
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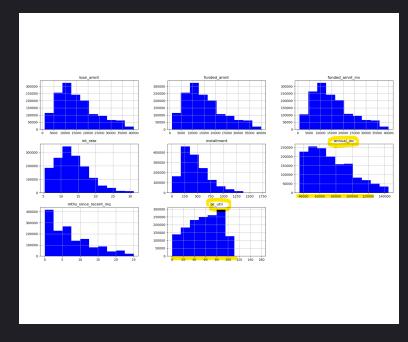
Big Data challenges

Univariate Analysis

Two major components: Outliers and Skewness







Outlier

can significantly distort the feature distribution and ML algorithms such as: Linear Regression, Logistic Regression, Support Vector Machine

Remove outlier

low_q = df.quantile(0.08)

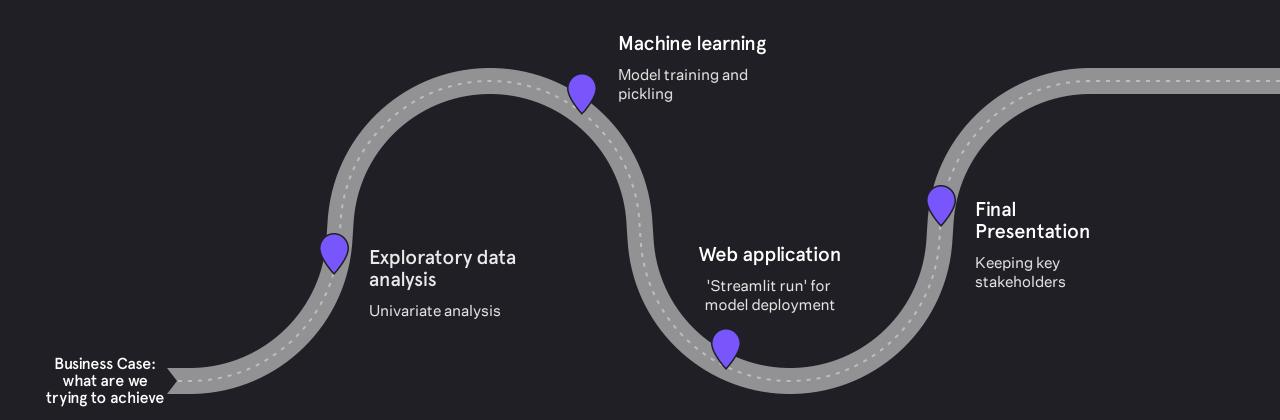
high_q = df.quantile(0.92)

 $# \text{ new_df} = (\text{df} < \text{q_hi}) & (\text{df} > \text{q_low})]$

Distribution after dropping outliers

If you have a really good sense of what range the data should fall in, like people's ages, you can safely drop values that are outside of that range.

Roadmap



Demos

