



# Exploring Borrower Characteristics and Loan Performance: A Lending Club Study

# Agenda

Dataset Introduction

Biases and Assumptions

Analysis

Recommendations

Final Thoughts

Q & A



# Dataset Introduction

Our dataset provides insight into Lending Club, a groundbreaking peer-to-peer lending firm headquartered in San Francisco, California.

Spanning from 2007 to around 2016, our dataset presents a comprehensive view of loans facilitated by Lending Club.

Lending Club offers a variety of loan products tailored to assist consumers with debt consolidation, weddings, and vacations.



# Objectives and Goals

Can we identify correlations between borrower credit and financial profiles to reduce the rate of charged-off loans, while also leveraging historical borrower characteristics to customize loan products and enhance customer retention?



## Goal #1

Exploring Charged off  
Loans



## Goal #2

Enhancing Customer  
Retention Rates



## Goal #3

Tailored Loan Products for  
Borrowers

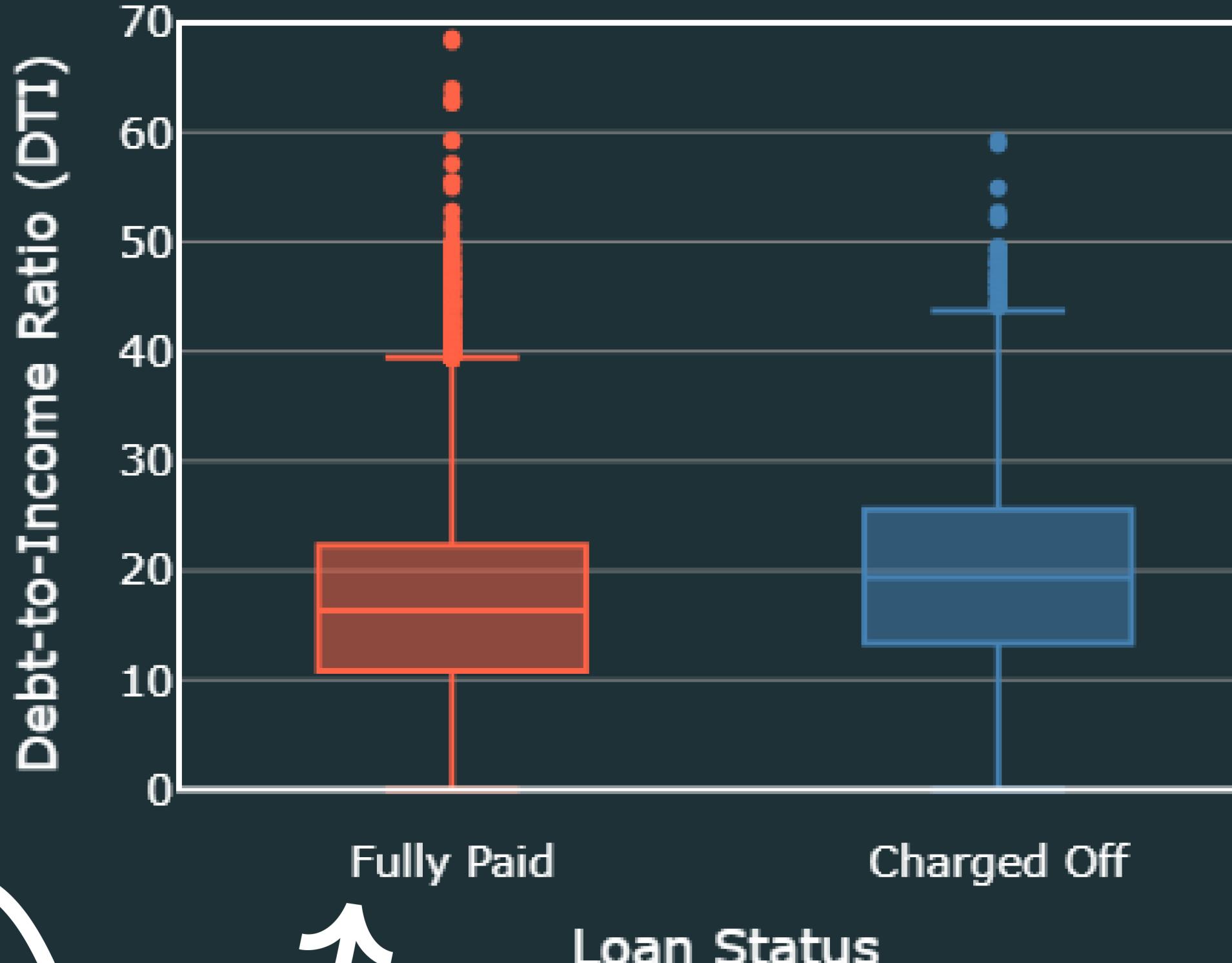
# Uncovering Dataset Biases and Assumptions

- Successful loans outnumber unsuccessful ones in the dataset.
- The dataset covers loans issued from 2007 to 2016.
- Annual income information has been grouped according to IRS 2024 income tax brackets, aiding in data retention and pinpointing areas for potential improvement
- Debt-to-Income (DTI) ratios have been limited to 70 to remove outliers that could skew statistical analysis.
- The dataset was obtained from Kaggle.



# Loan Status vs Debt-to-Income Ratio (DTI)

## Distribution of DTI by Loan Status



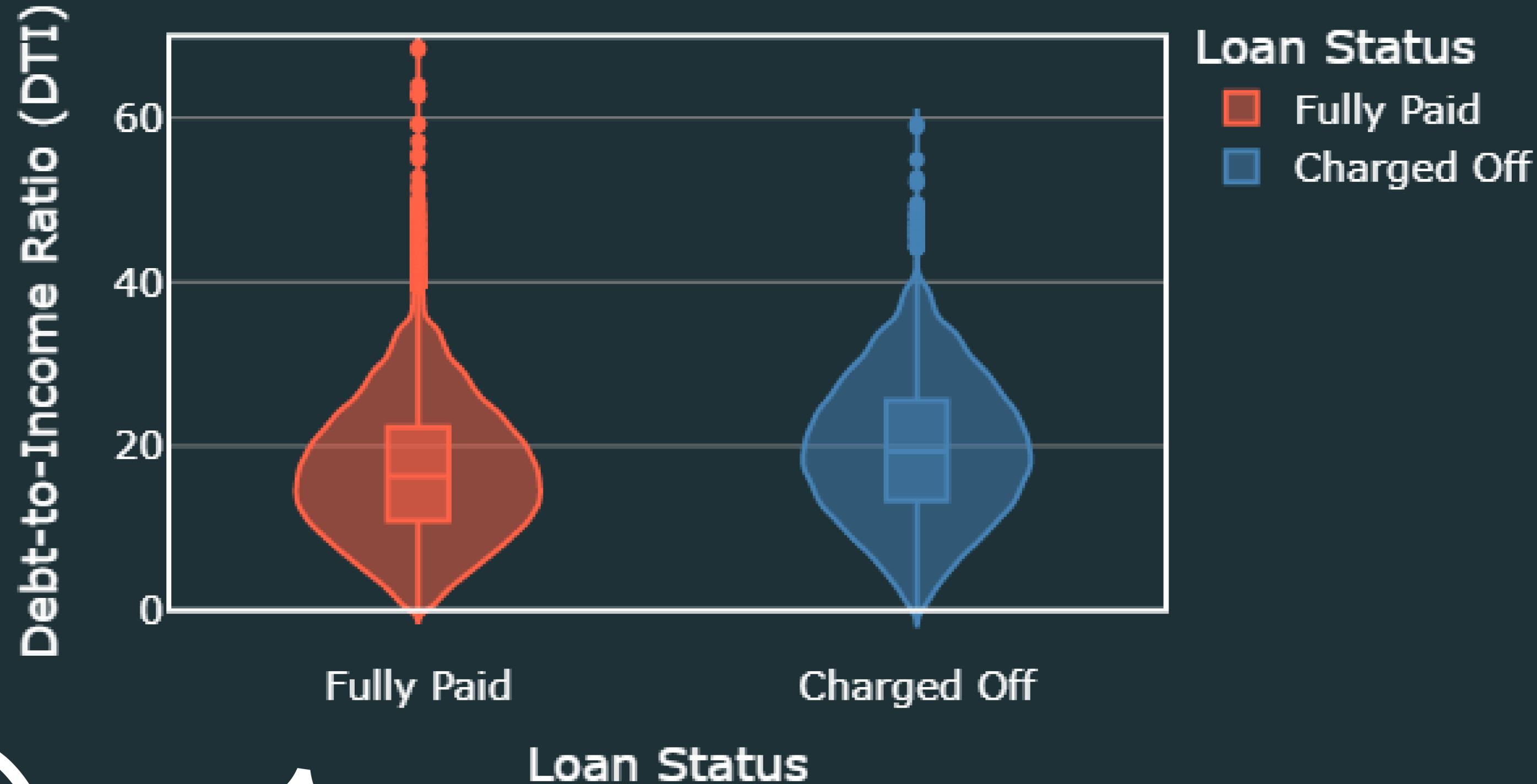
### Loan Status

- Fully Paid
- Charged Off

- Charged-off loans have higher median Debt-to-Income (DTI) ratios.
- Statistical tests confirm the DTI differences are significant.
- Further analysis shows a moderate impact, reinforcing that higher DTI is common in charged-off loans.

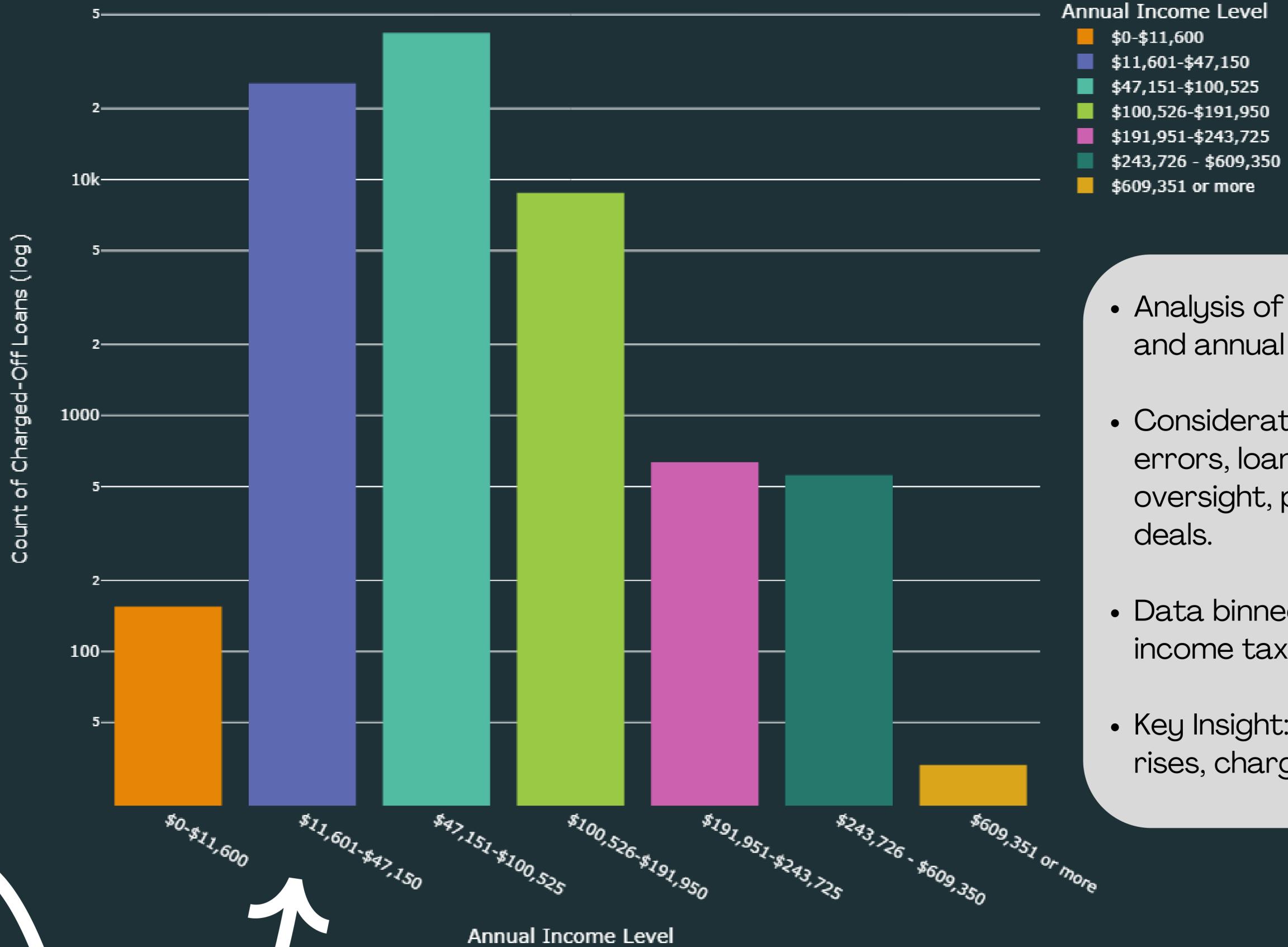
# Loan Status vs Debt to income Ratio (DTI)

## Distribution of DTI by Loan Status



# Loan Status vs Annual Income

## Distribution of Loans by Annual Income Level and Charged-Off Status



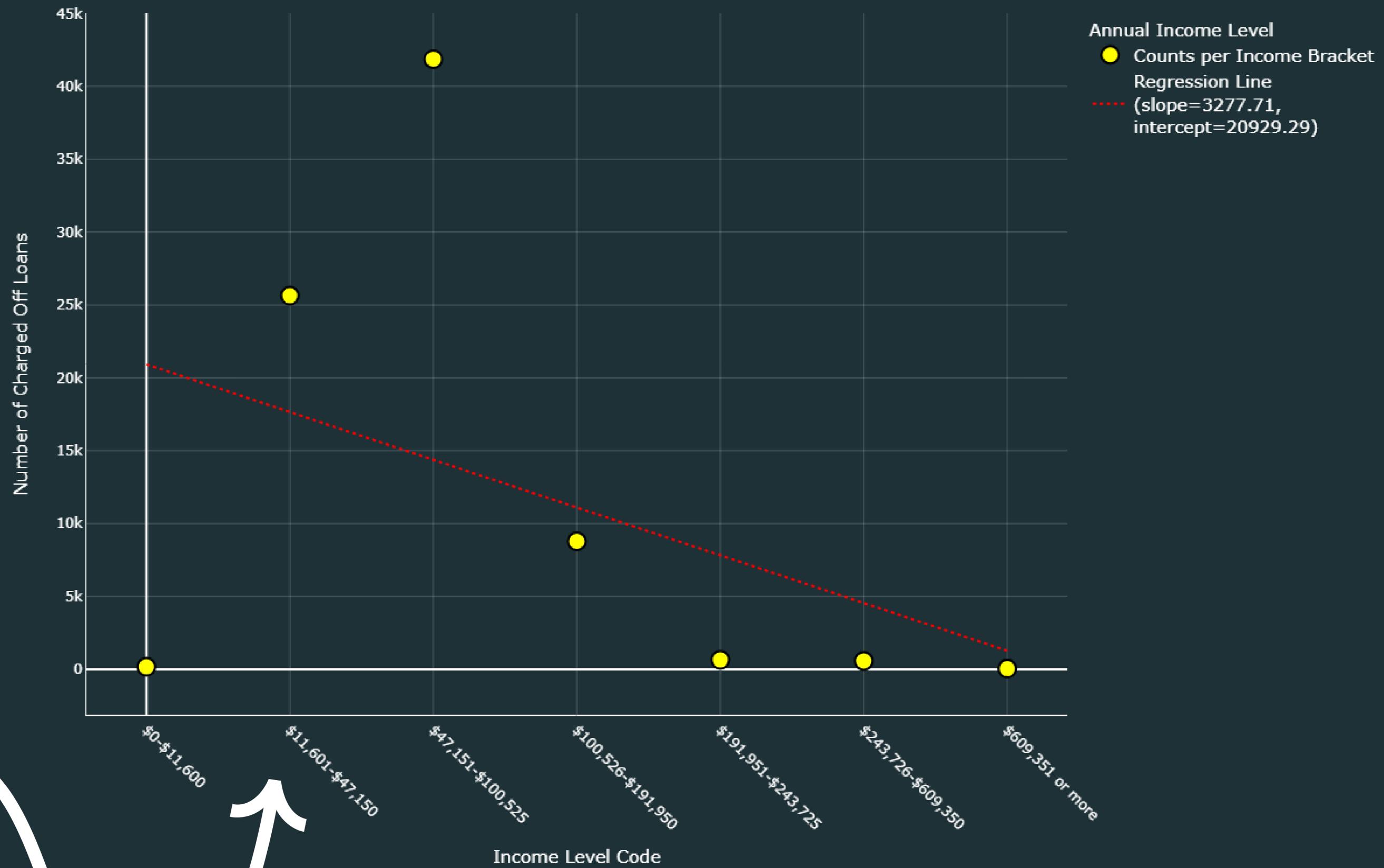
Annual Income Level

- \$0-\$11,600
- \$11,601-\$47,150
- \$47,151-\$100,525
- \$100,526-\$191,950
- \$191,951-\$243,725
- \$243,726 - \$609,350
- \$609,351 or more

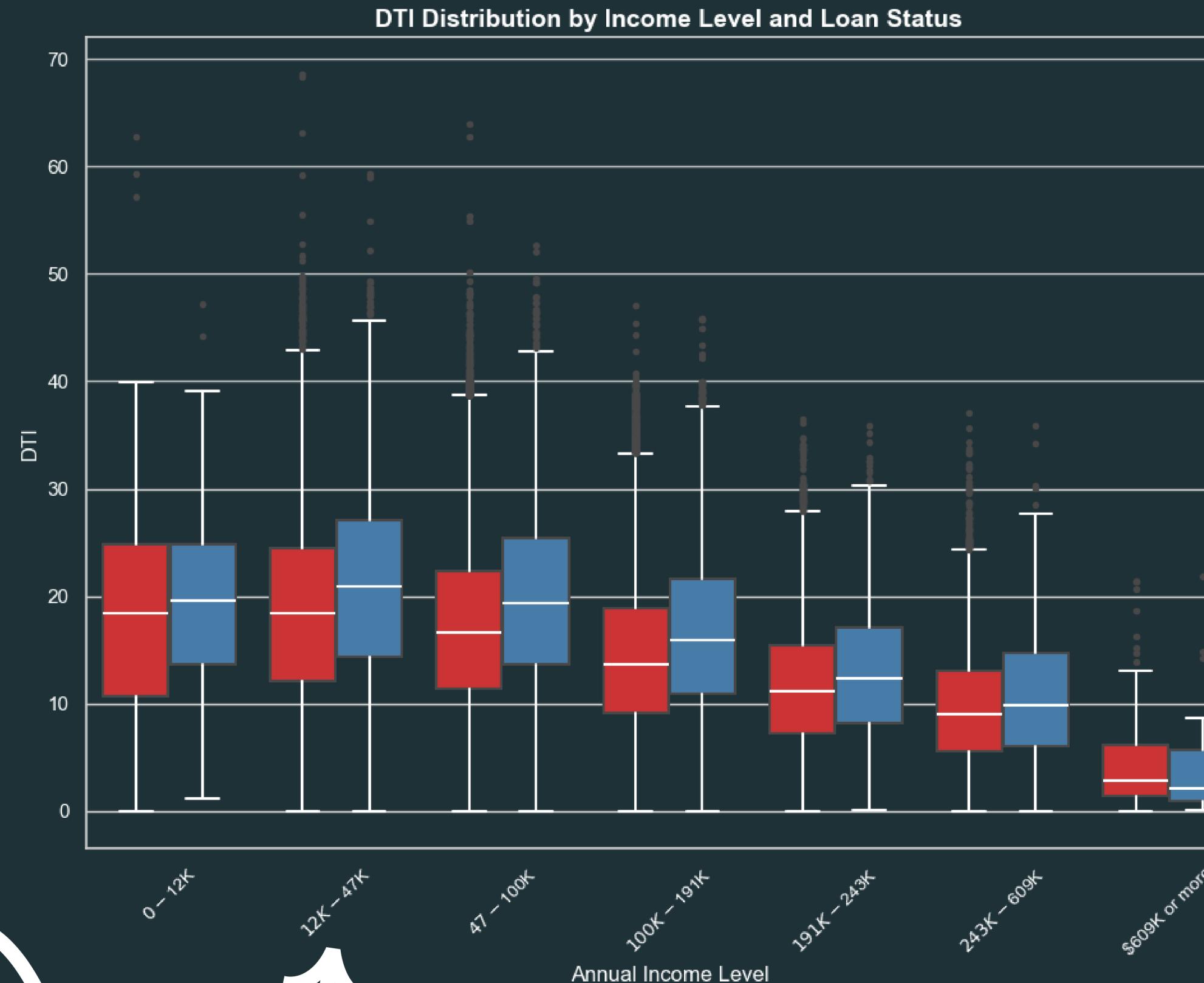
- Analysis of loan status and annual income.
- Considerations: borrower errors, loan officer oversight, potential dead deals.
- Data binned by 2024 income tax brackets.
- Key Insight: As income rises, charge-offs decline.

# Loan Status vs Annual Income

## Linear Regression Analysis of Charged Off Loans by Income Level



# Loan Status vs DTI & Annual Income



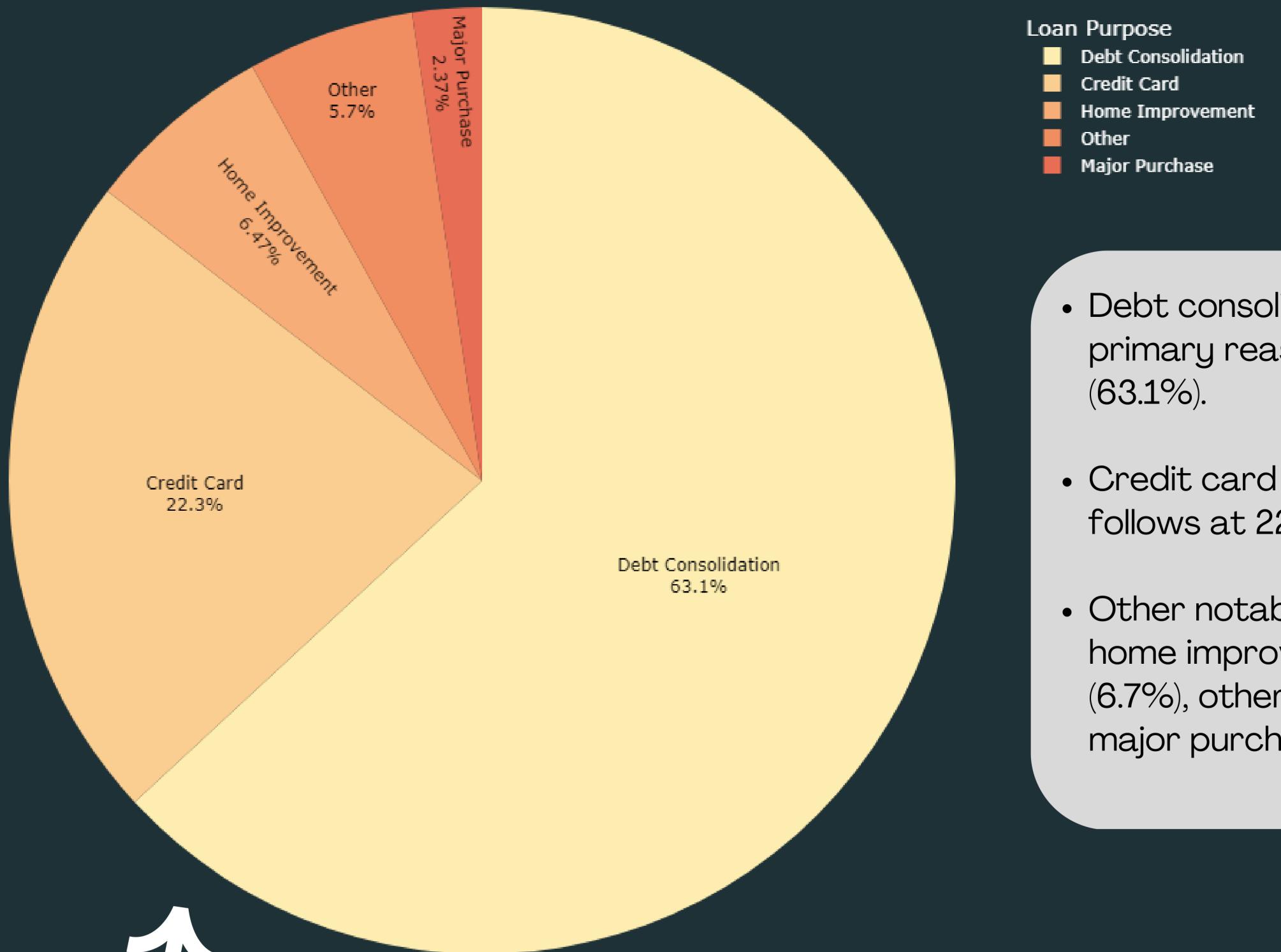
Loan Status

- Fully Paid
- Charged Off

- Multivariate analysis checks income bins with DTI grouped by charged-off loans.
- Higher DTI levels in charged-off loans across income bins.
- Box plot shows charge-offs have higher interquartile distributions than fully paid loans.
- Used DTI < 70 for consistency.

# Loan Purpose Analysis

**Distribution of Loans by Purpose**



**Loan Purpose**

- Debt Consolidation
- Credit Card
- Home Improvement
- Other
- Major Purchase

- Debt consolidation is the primary reason for loans (63.1%).
- Credit card repayment follows at 22.3%.
- Other notable purposes: home improvement (6.7%), other (5.7%), and major purchases (2.3%).

# Recommendations



# 1

**Marketing Strategy for  
Debt Consolidation**



# 2

**Implement Dynamic Risk  
Models**

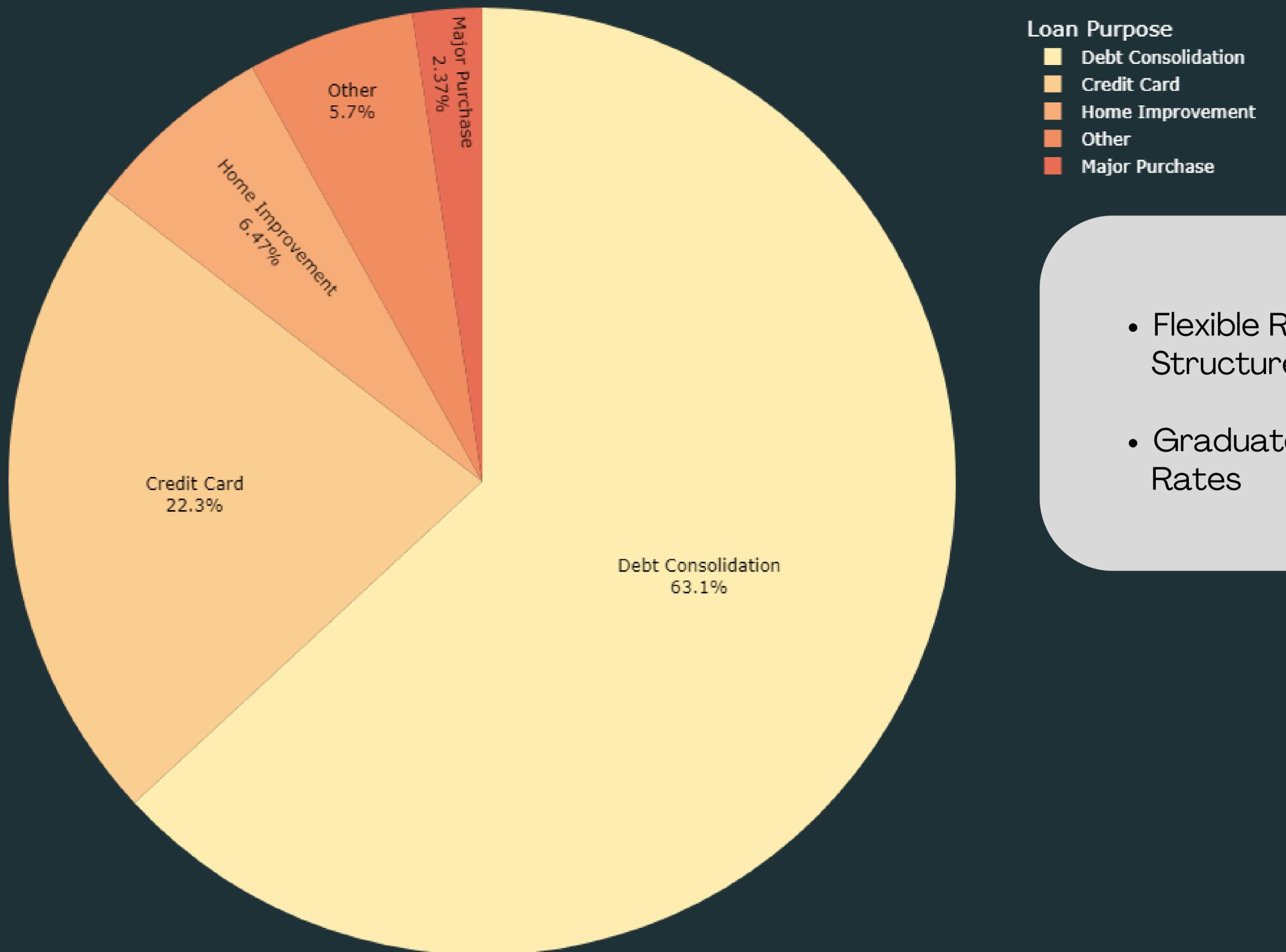


# 3

**Tightening Income  
Verification**

# Marketing Strategy for Debt Consolidation

## Distribution of Loans by Purpose



### Loan Purpose

- Debt Consolidation
- Credit Card
- Home Improvement
- Other
- Major Purchase

- Flexible Repayment Structures
- Graduated Interest Rates

# Implement Dynamic Risk Models

## Benefits

- Predict the probability of loan charge-offs
- Proactive risk management



# Tightening Income Verification

## Improve

- Data Integrity
- Financial Literacy Programs





# Final Thoughts & Future Directions





# Q & A

