credit score

purpose :developing a credit score model to assess credit risk and support informed loan approval decisions

result: Provided insights into the impact of different cutoff scores on acceptance rates, bad rates, and profitability, enabling informed decision-making for lenders.

methodology: model use <u>OptBinning</u> library in python to do all this



- bin each independent variable
- calculate each bin's information value (IV)
- calculate the log of (% of non-events / % of events) as the woe value
- replace each observation of the independent variable with the woe value
- perform WOE logistic regression on the dependent variable against the new woe dataset that has a monotonic trend in woe, and select variables based on IV
- evaluate model and create credit score

result of model

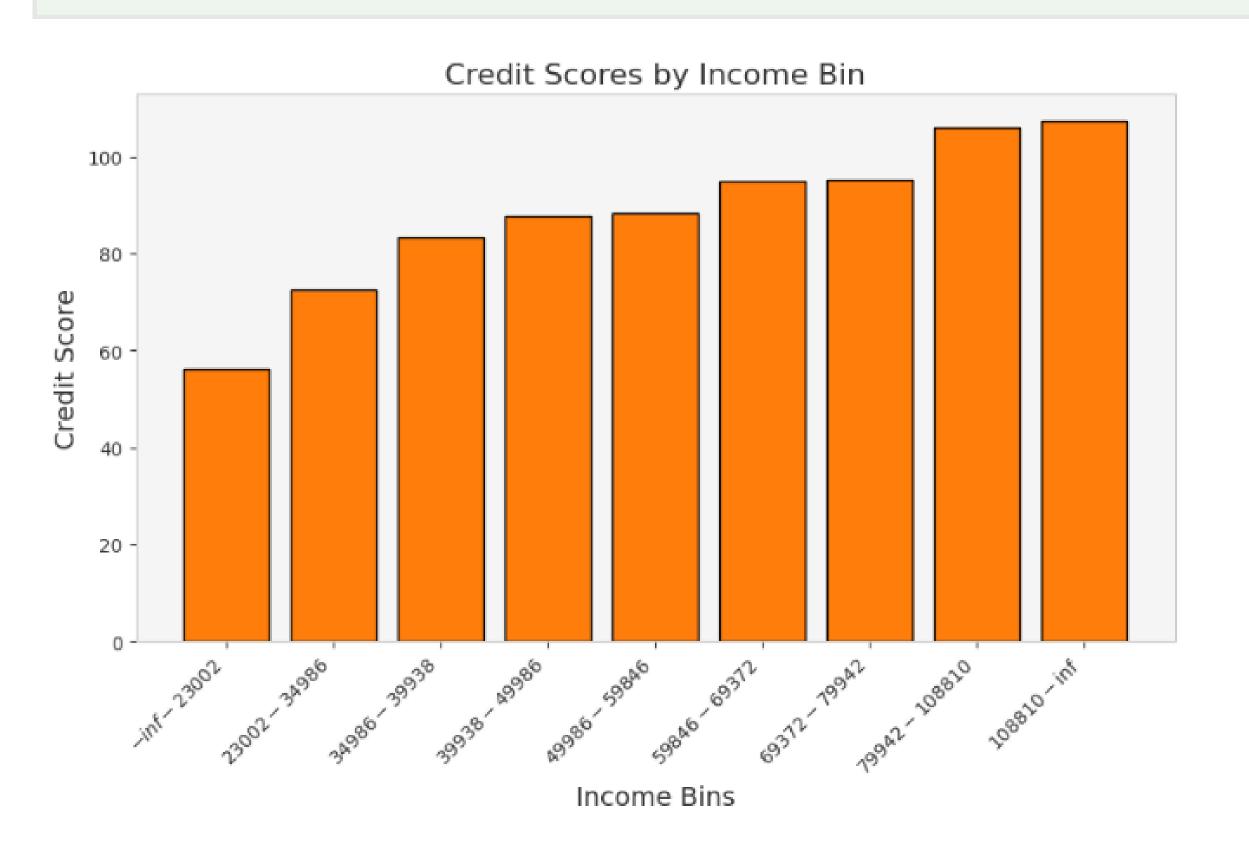


- Variable: person_income, Coefficient: -0.6266
- Variable: person_home_ownership, Coefficient: -0.8845
- Variable: person_emp_length, Coefficient: -0.2723
- Variable: loan_intent, Coefficient: -1.2654
- Variable: loan_percent_income, Coefficient: -0.9827
- Variable: cb_person_default_on_file, Coefficient: -1.1686
- auc = 0.82
- gini = 0.65

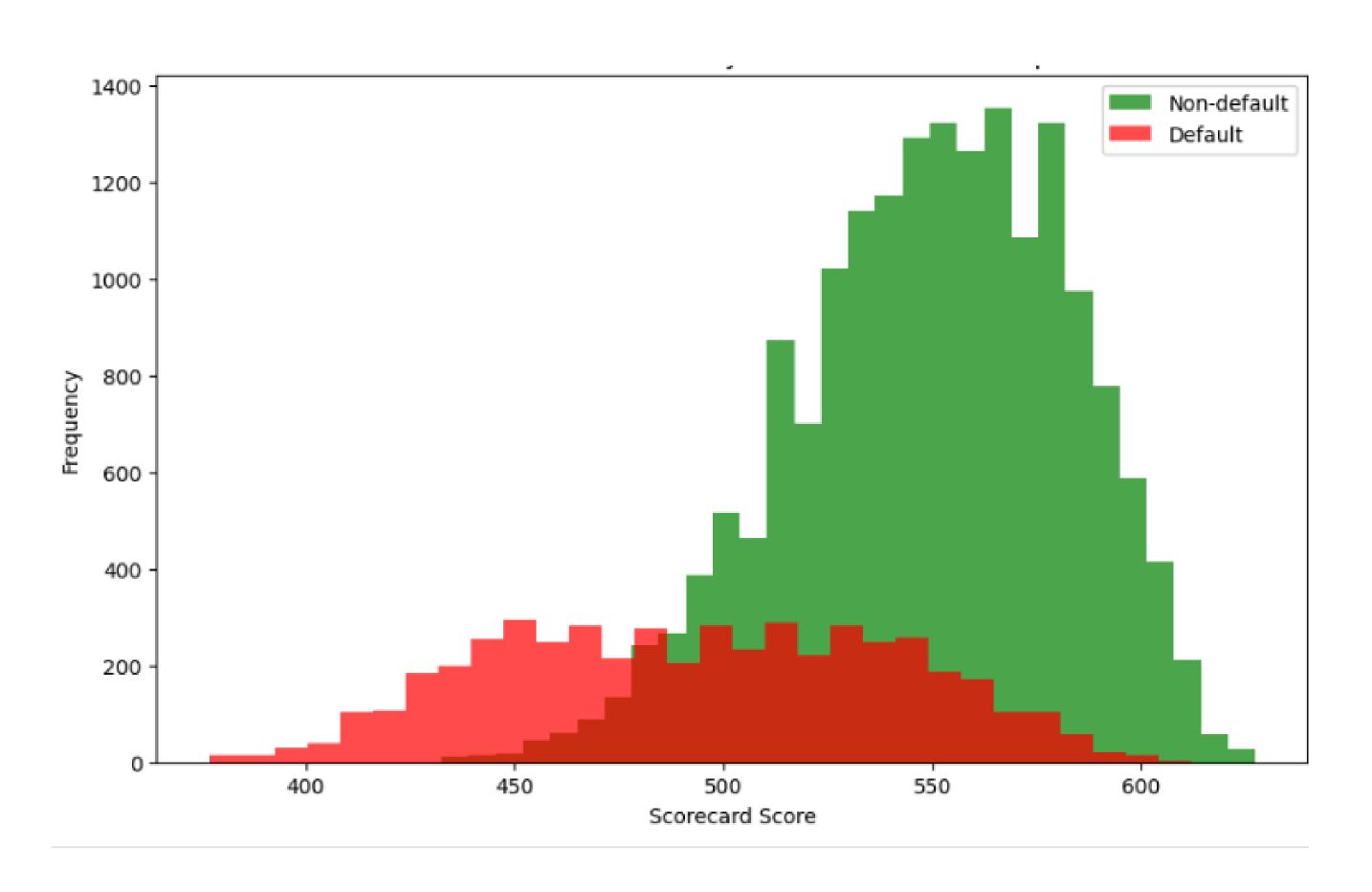
example of credit score

Higher income levels get higher score points

• as income increases, the risk of defaulting on a loan decreases.

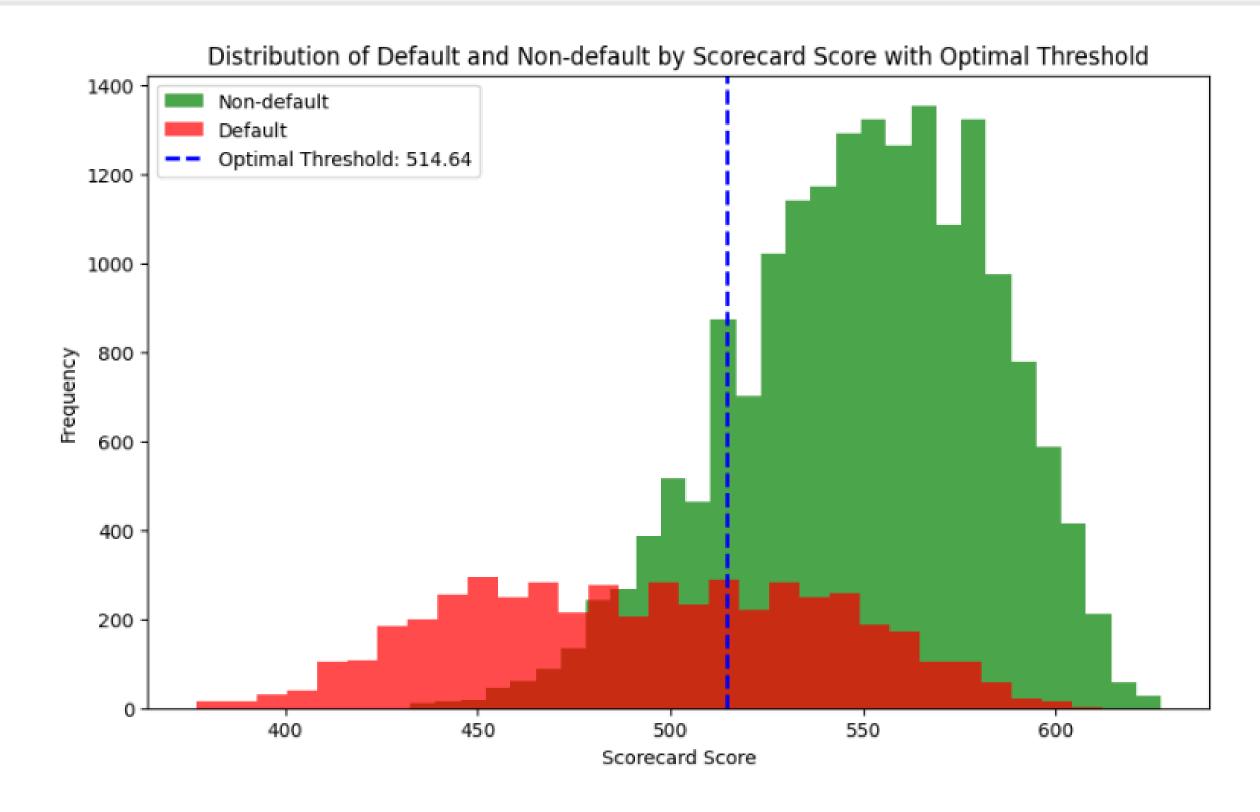


clear separation in score distributions



cut off score analyze: 1. maximizes the difference between the two distributions

- For each score bin, calculate the difference between the percentage of defaults and non-defaults
- The score value where this difference is largest represents the optimal cutoff point
- At this cutoff, you maximize the separation between the default and non-defaul



| | | Score Threshold | Cumulative Bad Rate | Cumulative Good Rate | Acceptance Kate |
|--|---|-----------------|---------------------|----------------------|-----------------|
| | 0 | 300 | 21.69% | 78.31% | 100.00% |
| | 1 | 400 | 21.44% | 78.56% | 99.75% |
| | 2 | 500 | 10.01% | 89.99% | 81.79% |
| | 3 | 514 | 7.98% | 92.02% | 74.34% |
| | 4 | 550 | 2.85% | 97.15% | 43.43% |
| | 5 | 600 | 0.05% | 99.95% | 3.63% |

and when we look at bad rate

- Cumulative Bad Rate represents the expected default rate if all loans with scores at given threshold are approved
- With a threshold of 514 (previously identified as optimal):
 - Strikes a balance between reasonable acceptance and manageable risk

| | loan_amnt | loan_int_rate | profit_loss | loan_status |
|-------|-----------|---------------|-------------|-------------|
| 0 | 35000 | 0.16 | -35000.00 | 1 |
| 1 | 1000 | 0.11 | 111.40 | 0 |
| 2 | 5500 | 0.13 | -5500.00 | 1 |
| 3 | 35000 | 0.15 | -35000.00 | 1 |
| 4 | 35000 | 0.14 | -35000.00 | 1 |
| | ••• | | | |
| 32576 | 5800 | 0.13 | 763.28 | 0 |
| 32577 | 17625 | 0.07 | 1320.11 | 0 |
| 32578 | 35000 | 0.11 | -35000.00 | 1 |
| 32579 | 15000 | 0.11 | 1722.00 | 0 |
| 32580 | 6475 | 0.10 | 646.85 | 0 |
| | | | | |

second method: profit

- (if loan does not default) = income from 1 year interest
- (if loan defaults) = negative value of principle

| | Score Threshold | Cumulative Bad Rate | Acceptance Rate | Profit |
|---|-----------------|----------------------------|-----------------|--------------|
| 0 | 350.00 | 21.82% | 100.00% | -54463425.11 |
| 1 | 400.00 | 21.57% | 99.76% | -53688950.12 |
| 2 | 450.00 | 17.19% | 95.22% | -35332310.50 |
| 3 | 500.00 | 10.09% | 81.83% | -9381296.51 |
| 4 | 514.64 | 7.86% | 73.74% | -5082928.43 |
| 5 | 537.00 | 4.62% | 56.13% | -125748.83 |
| 6 | 538.00 | 4.48% | 55.31% | 14842.43 |
| 7 | 550.00 | 2.85% | 43.50% | 2292758.42 |
| 8 | 600.00 | 0.04% | 3.59% | 767715.67 |
| 9 | 650.00 | 0.00% | 0.00% | 0.00 |
| | | | | |

538.0 score threshold:

- Optimizes profit at \$14,842.43
- 55.31% acceptance rate
- Prioritizes profitability and low-risk portfolio by accepting only higherscored applications
- A higher threshold would further limit the acceptance rate

strategy cut off: Prioritizes profitability and low-risk portfolio

| Score | Cumulative Bad | Acceptance | profit |
|-----------|----------------|------------|-----------|
| Threshold | Rate | Rate | |
| 538+ | 4.48% | 55.31% | 14,842.43 |

This approach may be suitable in the following scenarios:

- Economic Downturn or Recessionary Environment lenders may adopt a more conservative stance to mitigate potential losses
- Risk-Averse Lending Approach: Some lenders may have a lower tolerance for risk and prefer to maintain a high-quality loan portfolio

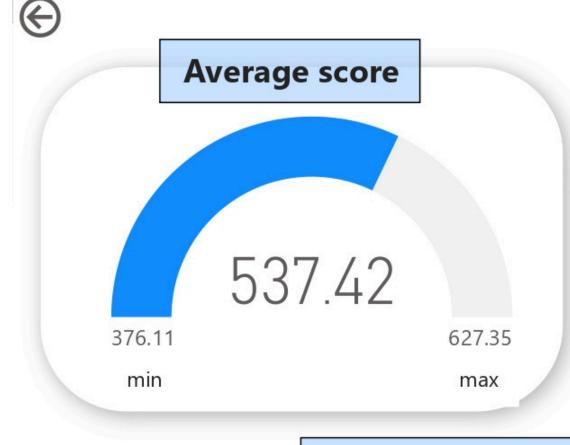
strategy cut off: Allows higher acceptance rate than 538 threshold, but with increased risk

| Score | Cumulative Bad | Acceptance | profit |
|-----------|----------------|------------|---------------|
| Threshold | Rate | Rate | |
| 514+ | 7.86% | 73.74% | -5,082,928.43 |

This approach may be advantageous in the following situations:

Aggressive Market Expansion: If the lender operates in a highly competitive market and seeks to gain a significant market share

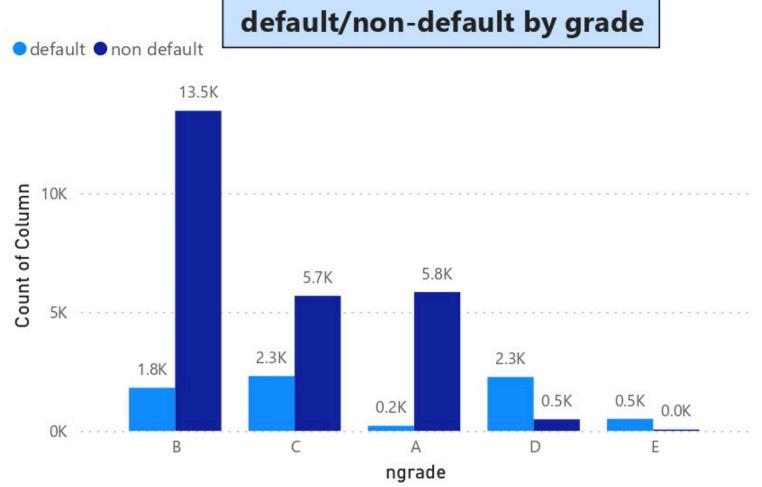
High Risk Tolerance: Lenders with a higher tolerance for risk may be willing to accept the potential losses



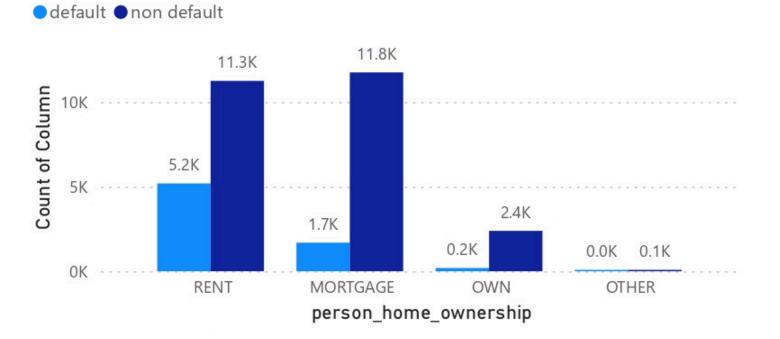
Sum of Loan Amounts

non default 235305925

default 77125375



Loan Count Comparison by Purpose



Median of person by grade

