SURVIVAL RATE ANALYSIS - UNSECURED PERSONAL LOAN PORTFOLIO

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SURVIVAL RATE MODELING

Analysis of Unsecured Personal Loan Portfolio Segemented by Credit Score, Rate and Original Loan Amount

This notebook demonstrates advanced survival analysis techniques using Kaplan-Meier and Cox regression models applied to unsecured personal loan portfolio data. We progress from exploratory data analysis through univariate survival modeling to risk-adjusted survival estimates and competing risk analysis. Lastly, we'll compile findings for credit risk management and portfolio optimization applications. Key Analyses:

- 1) Portfolio EDA and survival model readiness assessment
- 2) Kaplan-Meier survival estimation with censoring adjustments for young loans
- 3) Cox Proportional Hazards modeling with time-varying covariates
- 4) Credit risk management applications (PD modeling, portfolio segmentation, loss forecasting)

1.0 Loan Portfolio Sumamry and Exploratory Data Analysis

1.1 Generate Simulated Loan Portfolio

score_bucket

 Prime
 1560

 Super-Prime
 1407

 Near-Prime
 1204

 Subprime
 284

Name: count, dtype: int64

Loan data generated successfully.

Total Loans: 4,455

Date Range: 2021-02-01 to 2025-01-31

Years Covered: 5

Overview

This analysis utilizes a simulated personal loan portfolio designed to replicate realistic market conditions and lending patterns for survival analysis modeling. The dataset provides a robust foundation for examining loan performance, default behavior, and portfolio risk dynamics.

Loan Volume & Timing

- Total Loans Generated: 5,000
- Origination Period: January 1, 2019 June 30, 2025
- Volume Distribution: Strategically weighted to reflect market conditions
- Higher origination volumes during low-rate environment (2019-2022)
- Reduced origination volumes post-rate increases (2022-2025)

Loan Terms & Structure

- Principal Amount Range: \$1,000 \$50,000
- Maturity Terms: 3 to 8 years (variable by loan)
- Product Type: Unsecured personal loans with fixed rates and terms
- Amortization: Standard fixed-payment structure

Credit Risk Profile

- Credit Scores: Randomly distributed across realistic range (350-850)
- 6-Month Refresh Scores: Secondary credit score data point for risk monitoring
- Risk Segmentation: Natural distribution across credit quality spectrum

Rate Environment Stratification

The interest rate structure was strategically designed to mirror actual market conditions.

Pre-Rate Increase Period (2019 – April 2022):

- Lower base interest rates reflecting historical market conditions
- Higher loan origination volumes
- Inverse correlation maintained between credit scores and rates

Rate Increase Period (April 2022 – 2025):

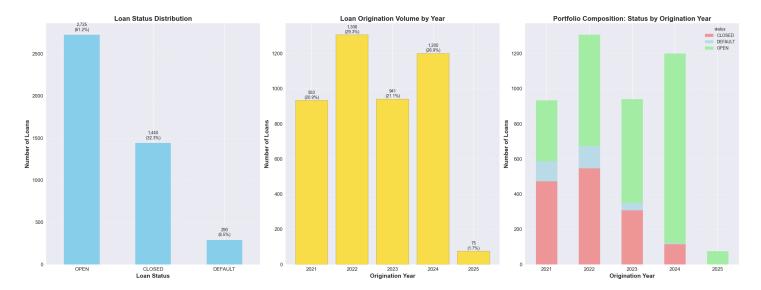
- Progressive rate increases of 5-6 percentage points
- Gradual implementation over 18-24 months
- Corresponding reduction in loan origination volumes
- Maintained credit-risk pricing differentials

1.2 Portfolio Summary and Review

Portfolio summary statistics results are the following:

Total Number of Loans: 4,455 Average Portfolio Rate: 12.33% Average Original Amount: \$13,905

Average Credit Score: 710



Maybe add % open, closed, default, total by year as a table

1.3 Key Metrics Summary by Credit Score, Rate, and Opening Amount

DEFAULT RATES BY ORIGINATION VINTAGE:

open_year	${ t Total_Loans}$	Defaults	Default_Rate_Pct	Avg_Duration_Months
2021	933	114	12.2	43.3
2022	1306	127	9.7	31.3
2023	941	44	4.7	18.8
2024	1200	5	0.4	7.3
2025	75	0	0.0	0.5

DEFAULT RISK ANALYSIS BY KEY SEGMENTS:

1. Default Rates by Credit Score Segment:

	${\tt Total_Loans}$	Defaults	Default_Rate_%
score_bucket			
Subprime	284	41	14.4
Near-Prime	1204	135	11.2
Prime	1560	90	5.8
Super-Prime	1407	24	1.7

2. Default Rates by Interest Rate Segment:

Total_Loans Defaults Default_Rate_%

Low	0	0	NaN
Low-Med	1	0	0.0
Medium	2351	156	6.6
Med-High	1610	93	5.8
High	489	40	8.2

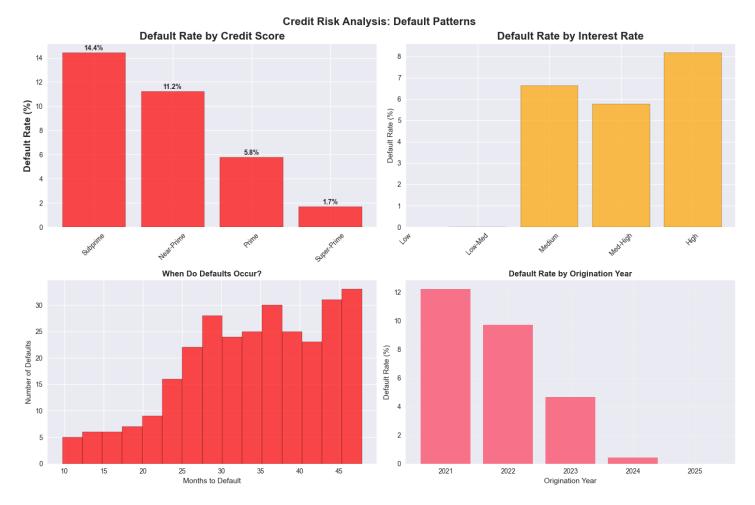
3. Risk Differentiation:

• Credit Score Risk Spread: 12.7 percentage points

Highest Risk Segment: 14.4% default rateLowest Risk Segment: 1.7% default rate

Add some commentary here

VISUALIZE DEFAULT RISK SEGMENTS:



Write a detailed summary about what I'm seeing in the data. May need to add another section on the Rate analysis

2.0 Kaplan-Meier Survival Rate Analysis

Add a brief writeup of the appraoch. Focus on defaulted. Binary targets: 1 = defaulted, 0 = closed or open. using months as the unit of measurement.

2.1 Baseline Kaplan-Meier Survival Rate and Cumulative Hazard Analysis

BASELINE SURVIVAL ANALYSIS - DATA PREPARATION:

Observation Date: 2025-01-31

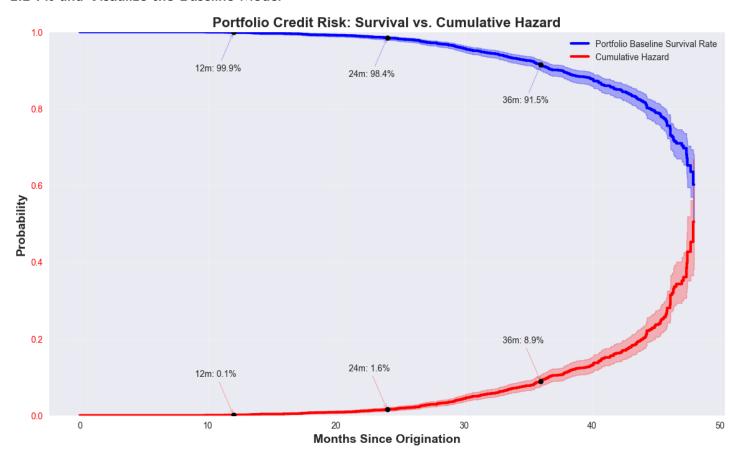
Total Loans: 4,455

Default Events: 290 (6.5%)

Censored Observations: 4,165 (93.5%)

Average Duration: 24.2 months
Duration Range: 0.0 to 48.0 months

2.2 Fit and Visualize the Baseline Model



Provide a summary

2.3 Evaluate Summary Statistics of Baseline Model

PORTFOLIO RISK METRICS:

Overall Default Rate: 6.5%

Average Observation Period: 24.2 months Estimated Annualized Default Rate: 3.2%

BASELINE SURVIVAL STATISTICS:

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6 months: 100.0% survival | 0.0% default rate | Cumulative Hazard: 0.0% 12 months: 99.9% survival | 0.1% default rate | Cumulative Hazard: 0.1% 18 months: 99.3% survival | 0.7% default rate | Cumulative Hazard: 0.7% 24 months: 98.4% survival | 1.6% default rate | Cumulative Hazard: 1.6% 30 months: 95.7% survival | 4.3% default rate | Cumulative Hazard: 4.4% 36 months: 91.5% survival | 8.5% default rate | Cumulative Hazard: 8.9%
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Median time to default (for loans that default): 34.6 months

Average time to default (for loans that default): 33.9 months

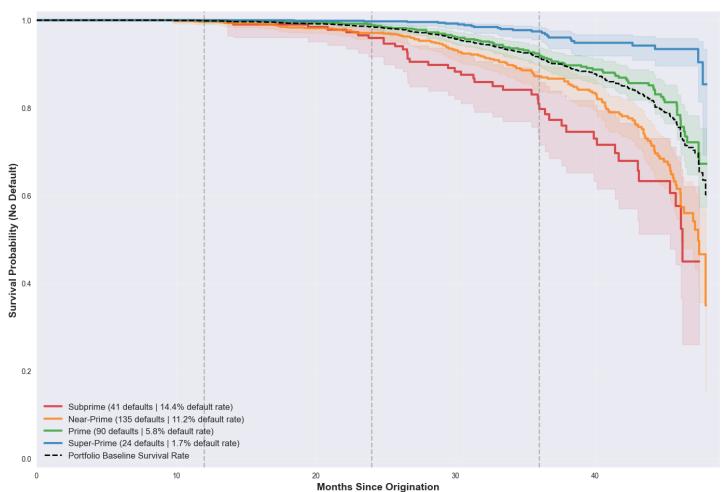
2.4 Fit and Visualize Models with Risk Segmentation

SEGMENTED CREDIT ANALYSIS:

	Risk Segment	Total Loans	Default Loans	Default Rate (%)	Average Duration
0	Subprime	284	41	14.4	24.18
1	Near-Prime	1204	135	11.2	24.97
2	Prime	1560	90	5.8	24.30
3	Super-Prime	1407	24	1.7	23.47

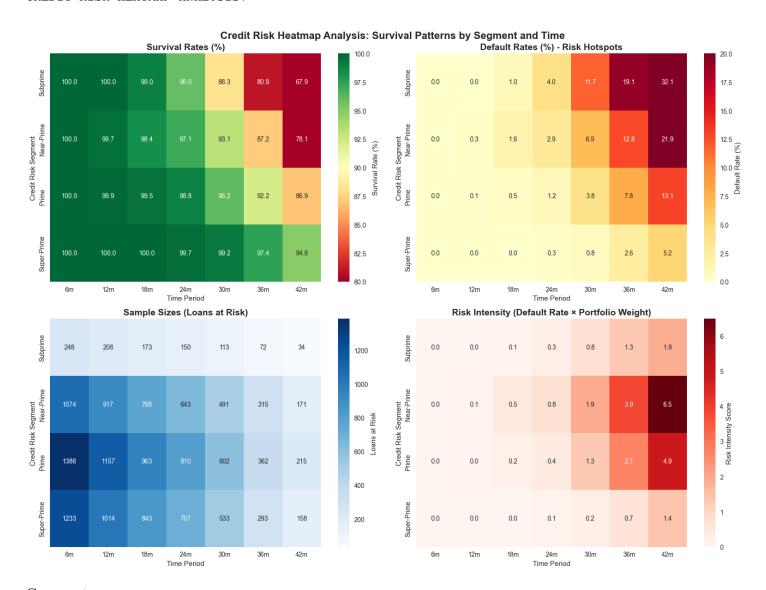
SEGMENTED SURVIVAL RATES:

Survival Analysis by Credit Risk Segment



Commentary

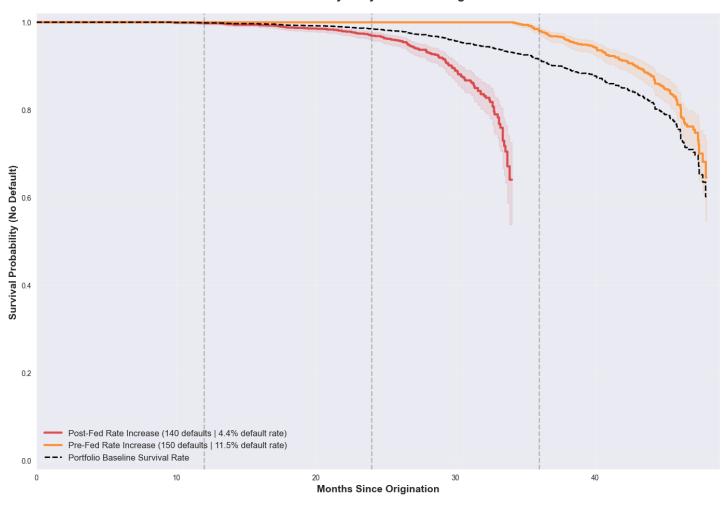
CREDIT RISK HEATMAP ANALYSIS:

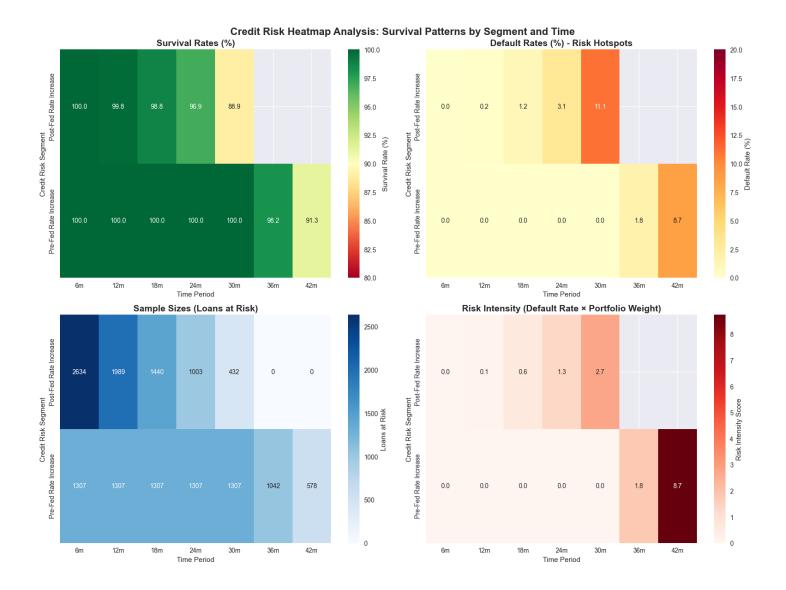


Commentary

2.5 Fit and Visualize Model with Interest Rate Period Segmentation

Survival Analysis by Credit Risk Segment





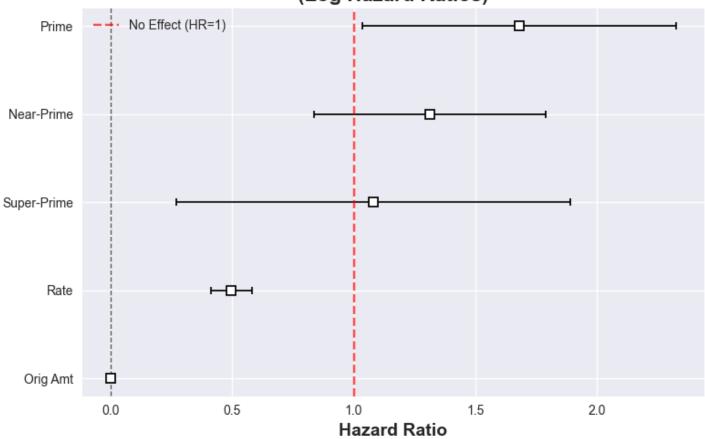
3.0 Cox Proportional Hazards Analysis

Commentry and approach

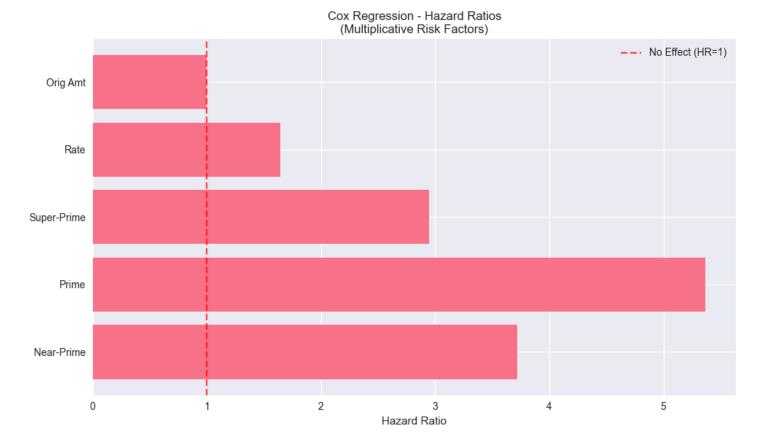
3.1 Fit and Assess the COX Model

COX REGRESSION COEFFICIENTS:

Cox Regression Coefficients (Log Hazard Ratios)



Interpretation: Positive coefficients indicate increased risk, negative coefficients indicate decreased



Discuss the paradox - subprime has highest default probability, but 1.00x hazard ratio due to a timing effect.

4.0 Risk Management Applications