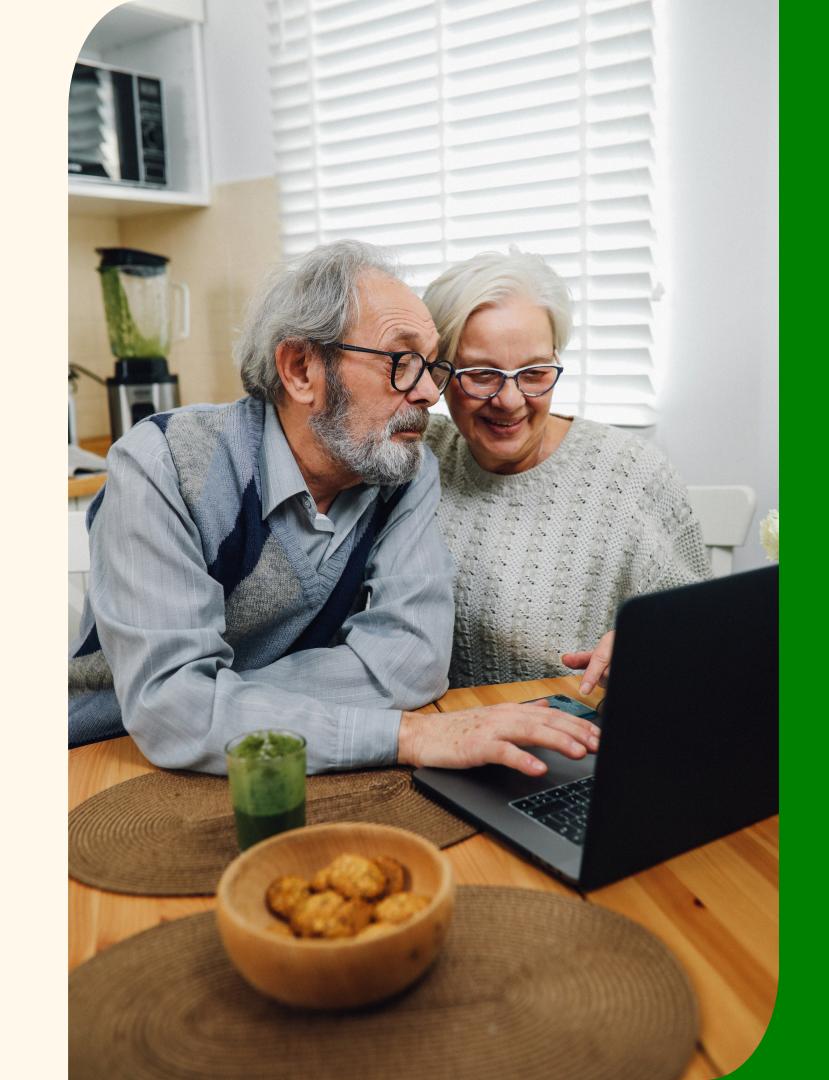
PREDICTIVE RISK MODELING FOR LIFE INSURANCE

Yenying Chen



OVERVIEW



CLEANING, HANDLING MISSING VALUES, STANDARDIZATION



FEATURE ENGINEERING, MODEL PERFORMANCE, EVALUATION METRICS



BUSINESS INSIGHTS, ACTIONABLE RECOMMENDATIONS, FUTURE ENHANCEMENTS

OBJECTIVES

- Improve risk assessment accuracy
- Optimize the underwriting process to increase the proportion of automated underwriting
- Reduce long-term claim risks caused by underwriting errors
- Balance risk management and premium revenue

DATASET

Prudential Life Insurance Assessment train: 59,381 train rows, 128 features test: 19,765 test rows, 127 features imbalanced dataset

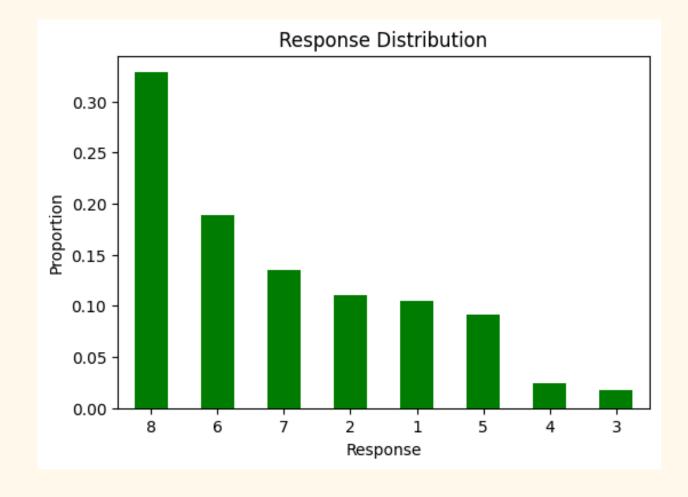
DATA CLEANING & FEATURE ENGINEERING

Data Cleaning

- Remove features with more than 50% missing values
- Fill missing values using median/mode
- Numerical Variables: Standardization
- Categorical Variables: Label encoding

• Feature Transformation and Engineering

- Add interaction features to strengthen relationships
- Merge multiple indicators to compute a composite risk score



MODEL PERFORMANCE & EVALUATION

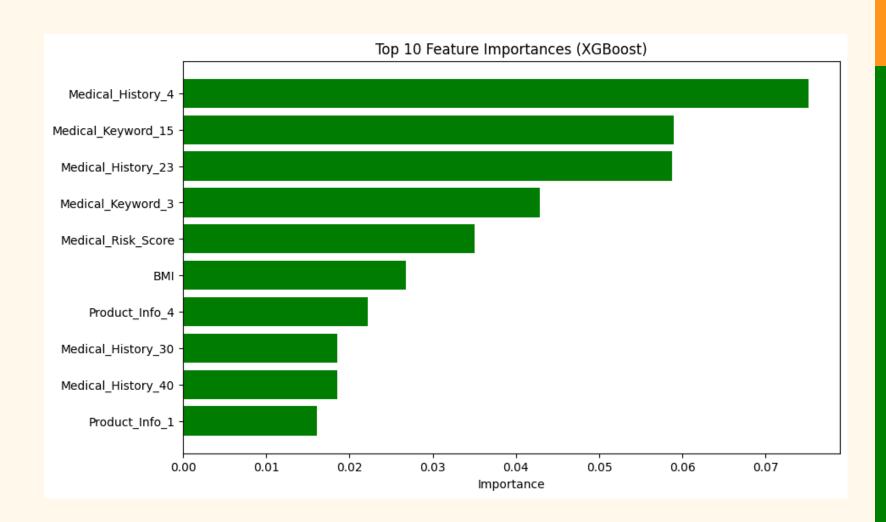
Metric / Model	Logistic Regression	Random Forest	XGBoost	XGBoost_best
Balanced Accuracy	0.286	O.318	O.381	0.380
ROC-AUC	0.801	0.825	0.842	0.847
Confusion Matrix – Recall (Classes 6–7)	71.13%	73.92%	73.44%	73.82%

MISCLASSIFICATION FINANCIAL IMPACT

Metric / Model	Logistic Regression	Random Forest	XGBoost	XGBoost_best
High-risk misclassified cases (Class 6-7 misclassified as 0-5)	3,766	3,367	3,727	3,740
Estimated future claims loss (\$50,000/person) (thousands)	\$188,300	\$168,350	\$186,350	\$187,000
Revenue loss due to customer churn (10- year LTV) (thousands)	\$4,033	\$4,614	\$3,647	\$3,790
Total financial impact (Claims + Churn) (thousands)	\$192,333	\$172,964	\$189,997	\$190,790

KEY RISK DRIVERS & BUSINESS APPLICATIONS

- Medical History is a key factor in underwriting decisions, highlighting the need for improved completeness and accuracy of health data.
- Using BMI and medical history analysis to enhance policyholder risk segmentation.
- Developing more refined premium adjustment strategies for different risk groups.
- Monitoring policyholder behavior patterns to improve fraud detection accuracy.



FUTURE OPTIMIZATION DIRECTIONS

Operational Applications & Continuous Monitoring

- Deploy models in underwriting systems for real-time risk assessment
- Dynamically adjust premium pricing and automated underwriting decisions based on risk evaluation results
- Flag high-risk cases to ensure manual review is more focused
- Establish data monitoring mechanisms, regularly adjusting decision thresholds to adapt to market changes (e.g., health data, medical trends)

Model Optimization

- Improve recall for low-risk customers to reduce wrongful underwriting rejections
- Apply ensemble learning to further optimize prediction capabilities
- Continuously fine-tune hyperparameters to enhance model accuracy
- Periodically retrain the model to maintain underwriting decision stability and financial control

THANKYOU

A&Q

APPENDIX: COMBINED RECALL FOR CLASS 6 & 7

WEIGHTED RECALL= (TP6+TP7)/(TP6+FN6+TP7+FN7)

- LOGISTIC REGRESSION
 (2013 + 17560)/(2013 + 2099 + 17560 + 1929) = 71.13%
- RANDOM FOREST
 (2192 + 18149)/(2192 + 2395 + 18149 + 1340) = 73.92%
- XGBOOST (3059 + 17148)/(3059 + 2096 + 17148 + 2341) = 73.44%
- XGBOOST_BEST (3056 + 17257)(3056 + 2053 + 17257 + 2232) = 73.82%

APPENDIX: FINANCIAL IMPACT CALCULATION

HIGH-RISK MISCLASSIFICATION (ACTUAL CLASS 6-7, PREDICTED AS 0-5)

- ACTUAL CLASS 6: 277 + 147 + 2 + 12 + 122 + 1,539 = 2,099
- ACTUAL CLASS 7: 149 + 81 + 2 + 17 + 48 + 1,370 = 1,667
- TOTAL MISCLASSIFIED CASES: 2,099 + 1,667 = 3,766
- COST CALCULATION: 3,766 × \$50,000 = \$188,300,000

LOW-RISK MISCLASSIFICATION (ACTUAL CLASS 0-1, PREDICTED AS 4-7):

- ACTUAL CLASS 0: 378 + 1,309 + 550 + 1,514 = 3,751
- ACTUAL CLASS 1: 809 + 1,521 + 512 + 1,472 = 4,314
- TOTAL MISCLASSIFIED CASES: 3,751 + 4,314 = 8,065
- CUSTOMER CHURN (10% CHURN RATE): 8,065 × 0.10 = 807
- CLV LOSS: 807 × \$500 × 10 = \$4,033,000

TOTAL: \$188,300,000 + \$4,033,000 = \$192,333,000

