Medical Device Recall Classification Implementation Report

5-Level Classification System

Generated on August 25, 2025

Data Science Team

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1. Dataset Overview

Total Records: 5000

Features: 8

Classification System: 5-Level Recall Severity

Training Set: 4000 records (80.0%) Validation Set: 1000 records (20.0%)

Stratified Sampling: Yes
Feather rocessing: Feature Striptiandization **Type**

text length Length of recall description Numeric

keyword count Medical keywords detected Numeric

severity score Automated severity calculation fumeric

has medical term Contains medical terminolog inary

risk score Calculated health risk score Numeric

complexity_score Technical complexity Numeric

urgency score Response urgency indicator Numeric

recall initiated Manufacturer-initiated recall Binary

Class Distribution: 5-Leyel Recall Classification (35.1%) Level 1: Critical (Death/Serious Injury) 1750 Level 2: Major (Temporary Health Issues) Level 3: Moderate (Minor Health Risk) Level 4: Minor (Field Safety Notices) 1511 Level 5: Information (Safety Alerts) (30.2%) 1500 1250 Number of Records 1000 739 (14.8%) 750 509 486 (10.2%) (9.7%) 500 250 0 Level 1 Level 2 Level 4 Level 5 Level 3

Feature Correlation Matrix

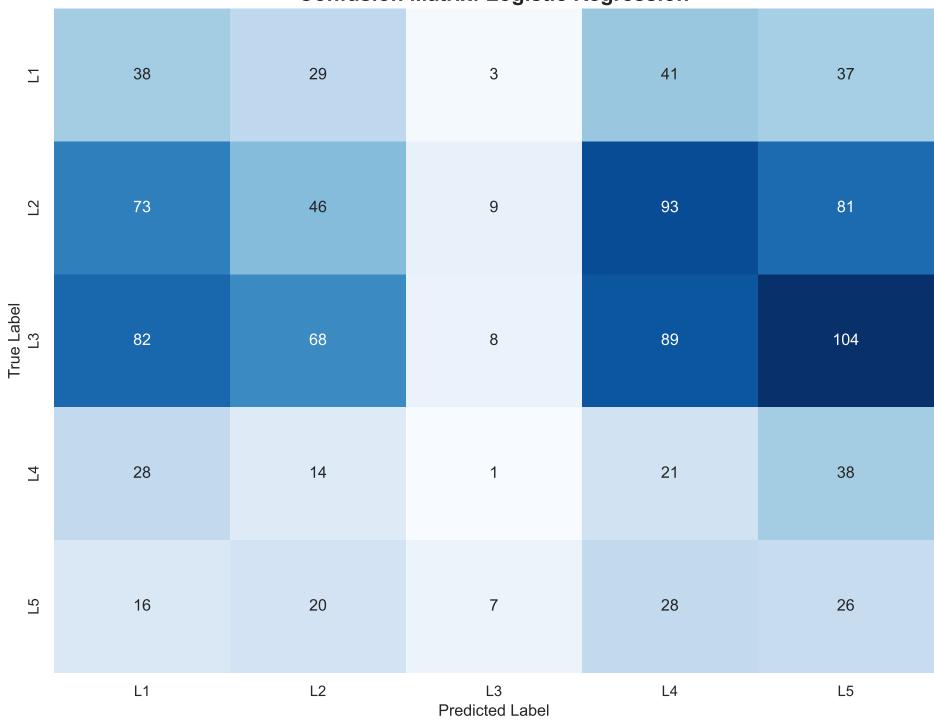
				Feature (Correlation	on Matrix					- 1.0
text_length	1.00	-0.02	0.02	0.00	0.01	-0.00	-0.01	-0.01	0.02		1.0
keyword_count	-0.02	1.00	-0.01	-0.01	0.00	0.01	-0.00	0.01	0.01		- 0.8
severity_score	0.02	-0.01	1.00	0.00	-0.02	0.00	0.01	0.00	0.00		
has_medical_term	0.00	-0.01	0.00	1.00	0.01	-0.02	0.01	-0.00	-0.02	-	- 0.6
risk_score	0.01	0.00	-0.02	0.01	1.00	-0.01	0.02	0.03	0.04		
complexity_score	-0.00	0.01	0.00	-0.02	-0.01	1.00	-0.01	-0.00	-0.02	-	- 0.4
urgency_score	-0.01	-0.00	0.01	0.01	0.02	-0.01	1.00	0.00	-0.03		
recall_initiated	-0.01	0.01	0.00	-0.00	0.03	-0.00	0.00	1.00	-0.00		- 0.2
recall_level	0.02	0.01	0.00	-0.02	0.04	-0.02	-0.03	-0.00	1.00	_	- 0.0
	text_length	keyword_count	severity_score	has_medical_term	risk_score	complexity_score	urgency_score	recall_initiated	recall_level		

Model Performance Comparison Training Accuracy Validation Accuracy 0.995 1.0 0.8 Accuracy 0.0 0.453 0.4 0.264 0.185 0.2 0.170 0.139 0.0 Logistic Regression Random Forest SVM

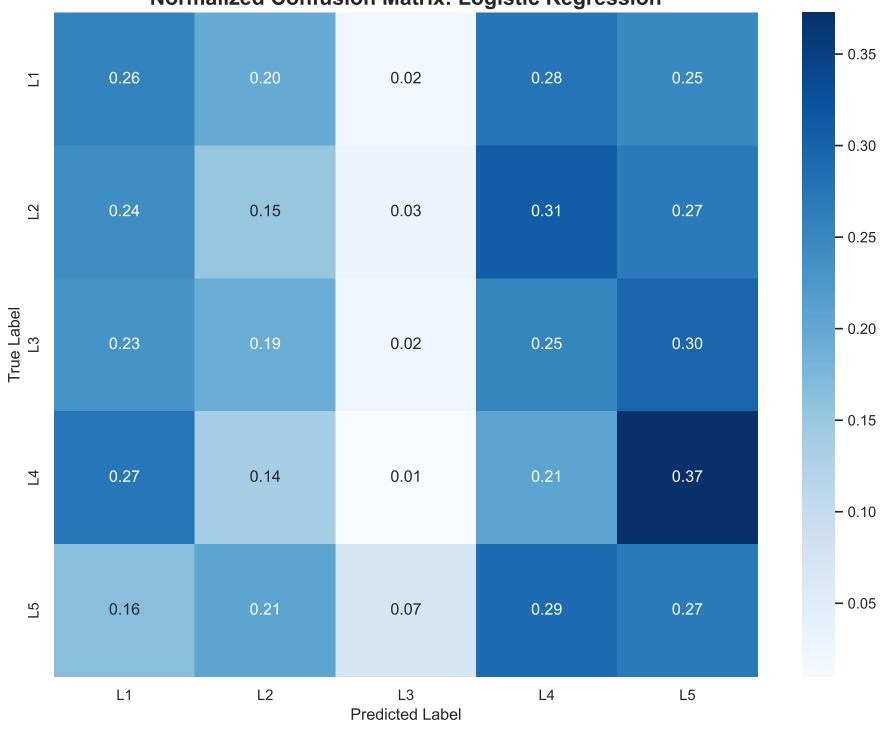
Models

Model Training Time Comparison 10.85s 10 Training Time (seconds) 6.00s 0.02s 0 Random Forest Logistic Regression SVM

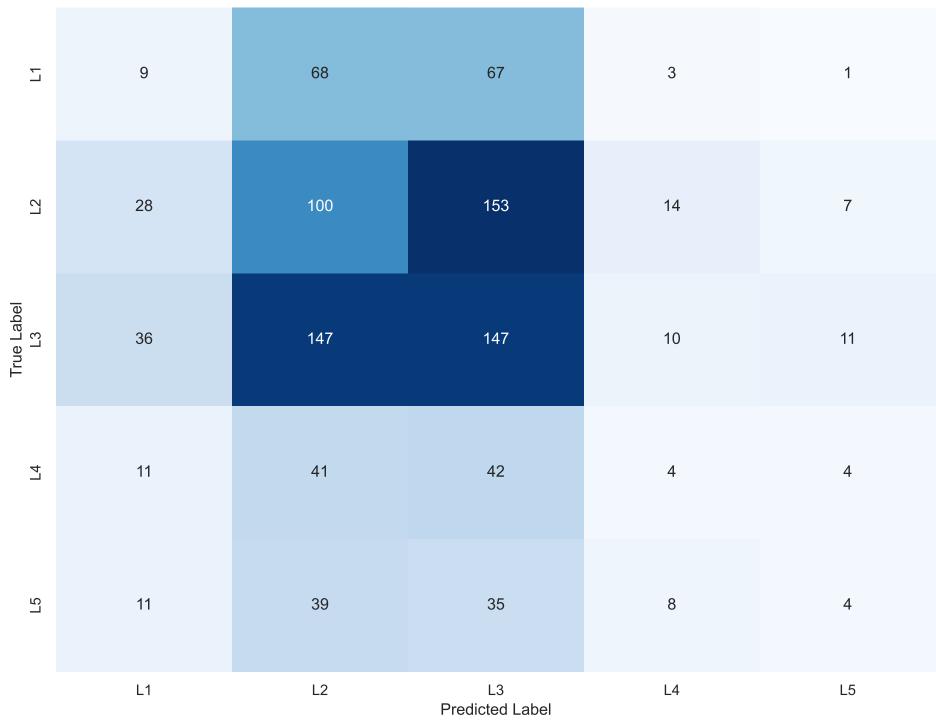
Confusion Matrix: Logistic Regression



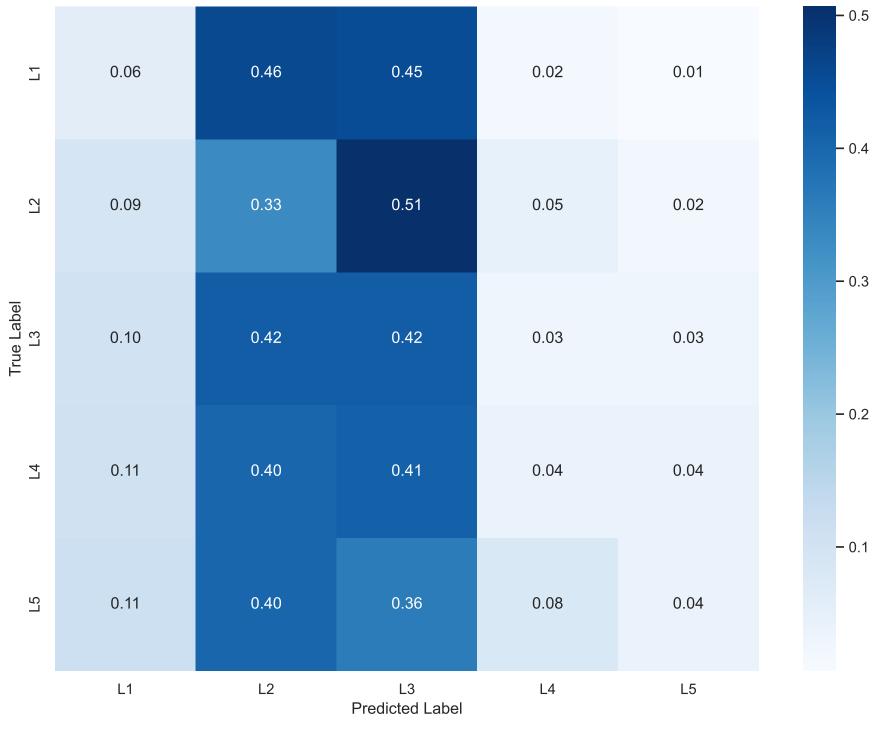
Normalized Confusion Matrix: Logistic Regression



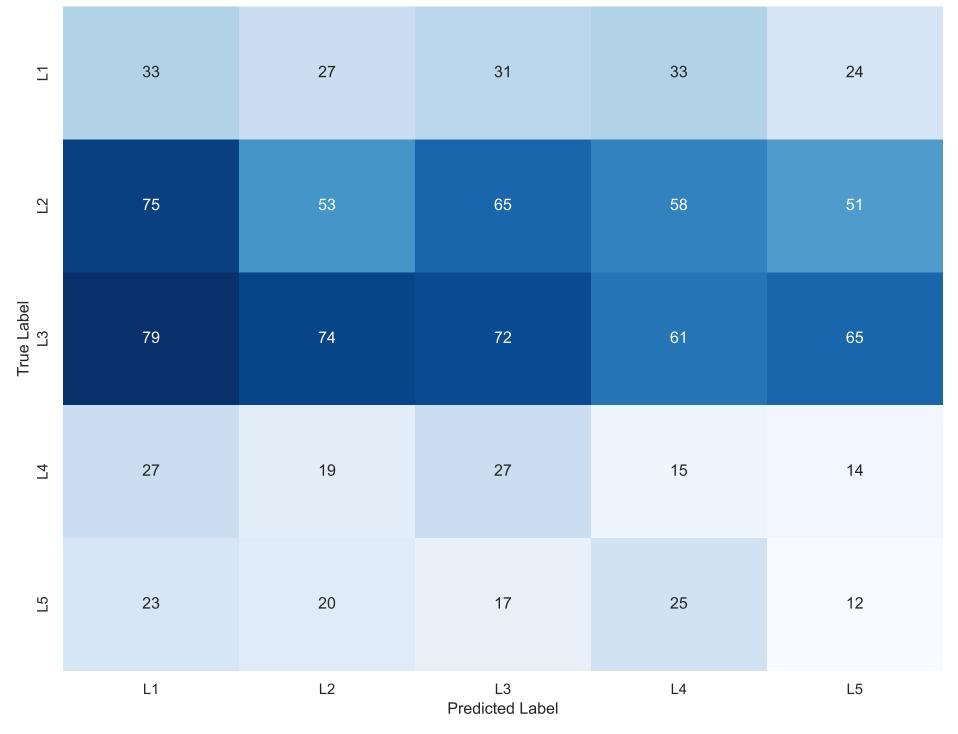
Confusion Matrix: Random Forest



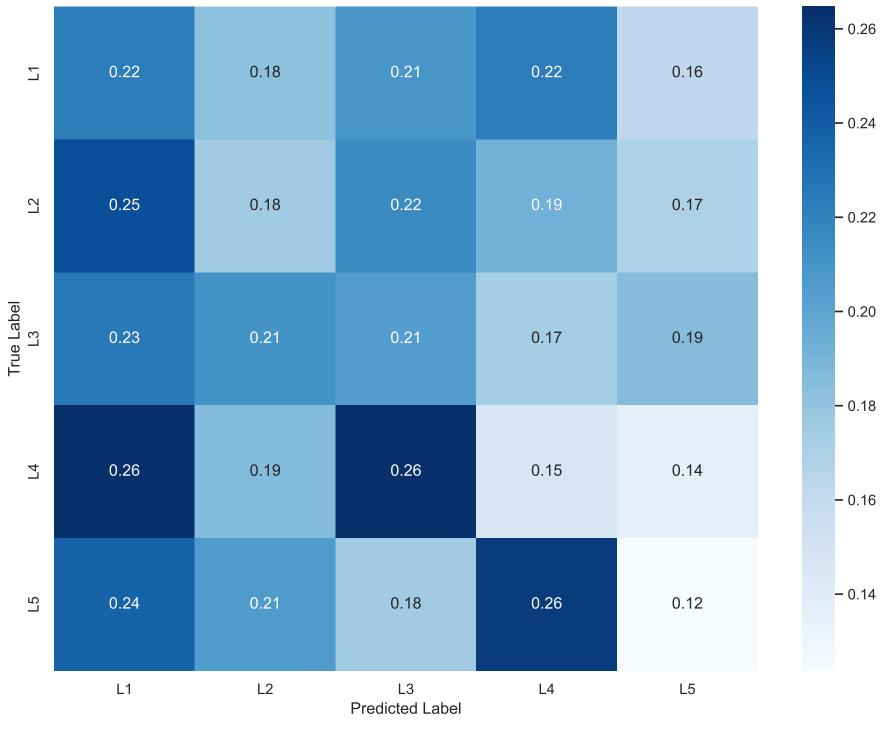
Normalized Confusion Matrix: Random Forest

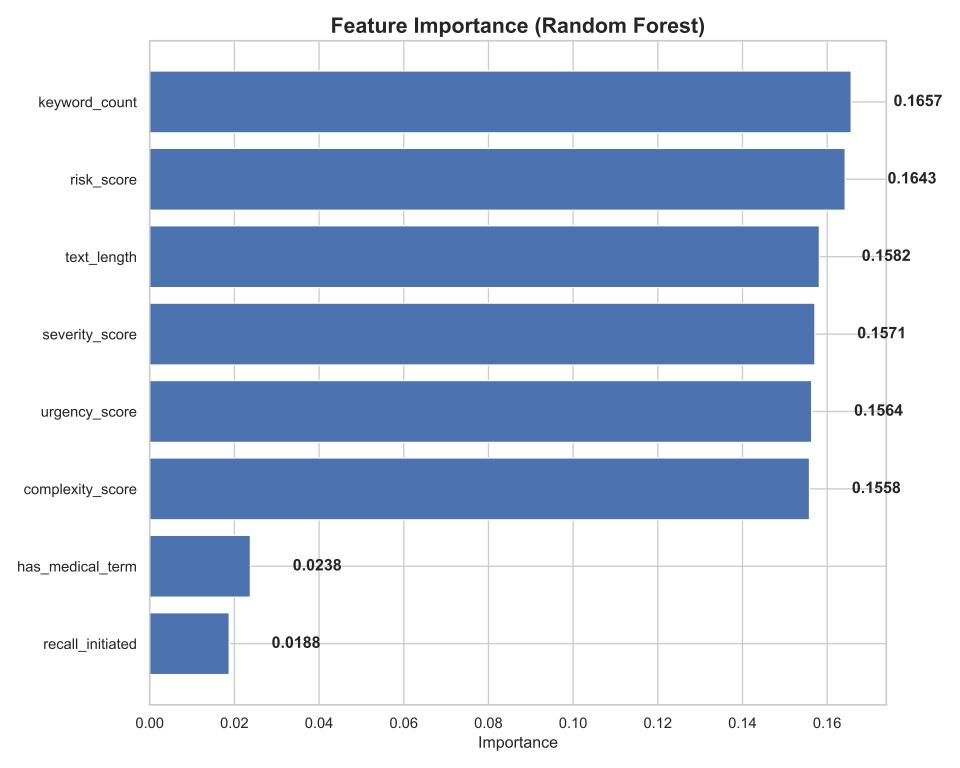


Confusion Matrix: SVM

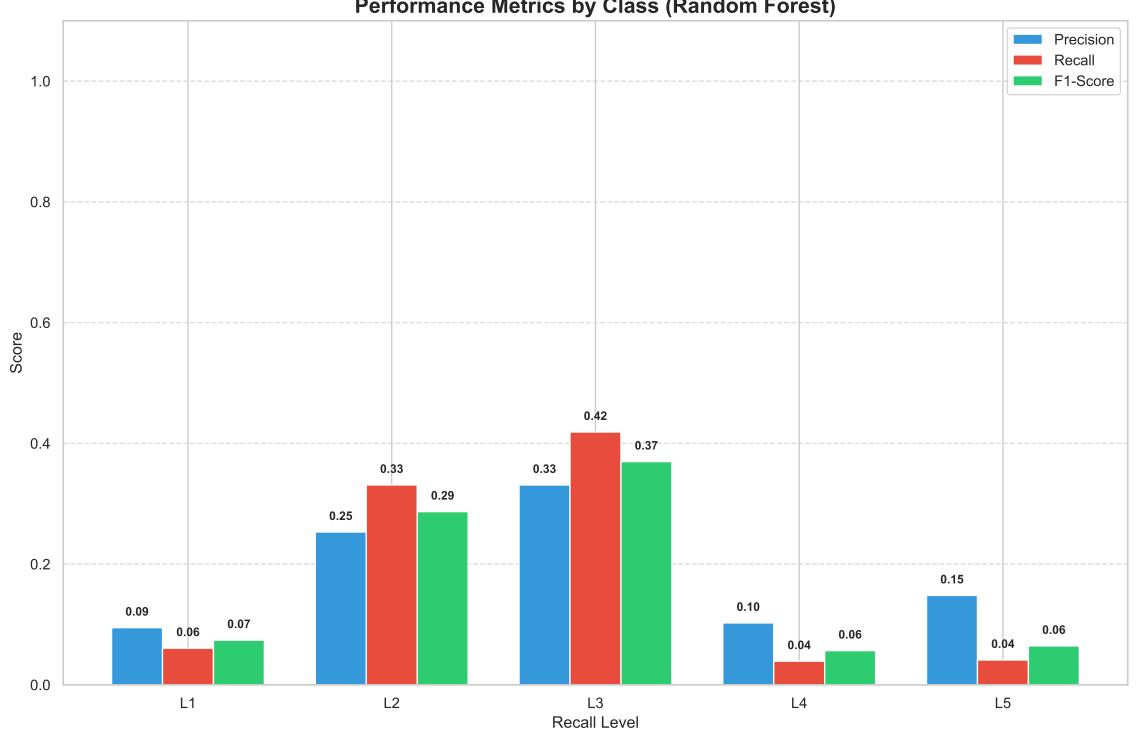


Normalized Confusion Matrix: SVM





Performance Metrics by Class (Random Forest)



Challenges & Limitations

1. Class Imbalance:

- Unequal distribution of samples across recall levels affects model training
- Balanced class weights were used to mitigate this issue

2. Feature Interpretability:

- Some features have complex relationships that are difficult to interpret
- Trade-off between model complexity and interpretability

3. Validation Limitations:

- No external validation dataset was available
- Cross-validation provides an estimate of performance, but real-world testing is needed

4. Multi-Class Complexity:

- 5-level classification is inherently more difficult than binary classification
- Some classes show confusion with adjacent levels (e.g., Level 2 vs Level 3)

5. Feature Engineering Opportunities:

- Current features could be enhanced with domain-specific medical knowledge
- Text-based features from recall descriptions are not fully utilized

Next Steps & Recommendations

1. Model Improvements:

- Hyperparameter optimization using grid search or Bayesian methods
- Ensemble methods combining multiple models for better performance
- Explore transformer-based models for text processing

2. Feature Enhancement:

- Incorporate text embedding techniques for recall descriptions
- Add domain-specific medical features based on expert knowledge
- Explore manufacturer history and reputation as features

3. Validation Protocol:

- Implement expert validation for critical classifications (Level 1)
- Create feedback loop for model improvement based on expert corrections
- Test on temporally separated data to evaluate real-world performance

4. Deployment Strategy:

- Develop confidence scoring mechanism for classifications
- Implement model monitoring for performance drift
- Create interpretability tools for regulatory stakeholders

5. Periodic Retraining:

- Schedule regular model retraining with new recall data
- Monitor feature distribution changes over time
- Update feature engineering pipeline as needed

Model Performance Summary

Metric	Logistic Regression	Random Forest	SVM	
Training Accuracy	0.1703	0.9950	0.4527	
Validation Accuracy	0.1390	0.2640	0.1850	
Training Time (s)	0.02	6.00	10.85	
Level 1 Recall	0.2568	0.0608	0.2230	
Level 2 Recall	0.1523	0.3311	0.1755	
Level 3 Recall	0.0228	0.4188	0.2051	
Level 4 Recall	0.2059	0.0392	0.1471	
Level 5 Recall	0.2680	0.0412	0.1237	
Macro F1-Score	0.1360	0.1704	0.1669	

