

Automated Tech Support Triage Model

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Automated Tech Support Triage Model

User Case

Classifies customer support tickets, identifies subsystem issues, and determines severity levels automatically.

Why Important

Saves time and resources, speeds up resolution, prioritizes critical issues, routes to right experts.

The Challenge

Limited training data, varied staff experience, manual classification slows identification across diverse systems.

Problem Statement & Approach



Our Goal

Automatically analyze report tickets to predict component origin of the issue and severity level.



Inputs

- User reported issues
- Reporter metadata
- System description



Outputs

- Component/subsystem location
- Severity: Low, Medium, High

What Makes This Novel

Dual Prediction

Predicts both component and severity, not just single label.

Enhanced Training

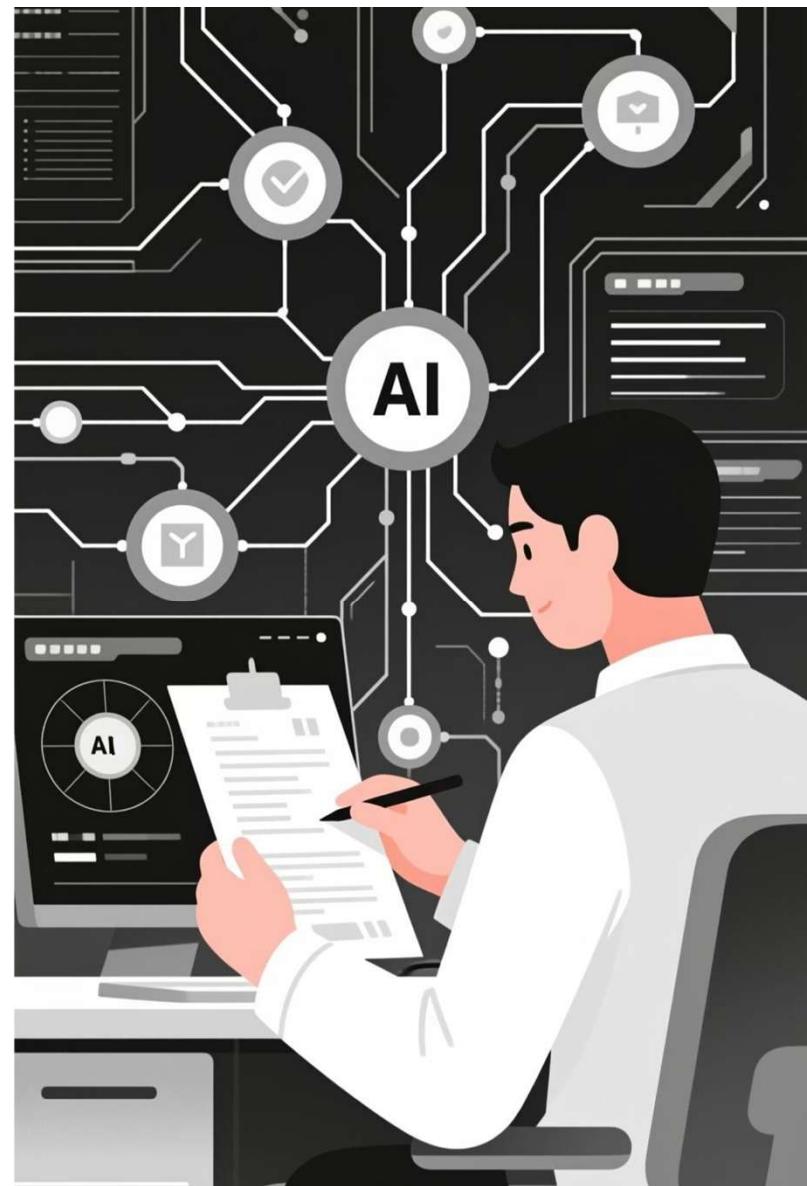
Uses multiple methods and larger training datasets than current systems.

Metadata Intelligence

Leverages reporter metadata to gauge reliability and differentiate real issues from user errors. (E.g. experience, role...)

Business Impact Focus

Severity based on domain-specific impact, not generic urgency.





Models & Techniques

Models

- Distilbert-base-uncased
Fine-tuned for ticket analysis
- Llama-3-8B
Fine-tuned for classification

Techniques

- Supervised Fine-Tuned Classifier
For severity determination
- Zero Shot Classification
For component attribution

Fine-Tuning Approach



Attribution Classifier

Build a labeled dataset and evaluate on held-out test data

Severity Classification

Cross-entropy classification loss



Dataset Generation Strategy

01

Create Synthetic Dataset

Simulate typical IT system errors with generated workflow and components

02

Generate Issue List

Develop possible issues for each system component

03

Create User Reports

Generate natural-sounding reports with severity and component attribution

Evaluation & Quality Metrics

Supervised Fine-Tuned Classifier



Data Split

70% train, 15% validation, 15% test



Train & Validate

Use labelled reports, validate to avoid overfit



Test Evaluation

Compute measurements on test set

Zero-Shot Classification



Prepare Dataset

Build labelled dataset for fine-tuning



Consistent Prompting

Run tests individually or batch



Held-out test evaluation

Manual vs predicted labels for component

Key Performance Indicators



Accuracy

Overall correctness



Precision

True positive rate



Recall

Coverage measure



F1 Score

Balanced metric



Confusion Matrix

Detailed breakdown



Macro-F1

Average F1 across classes