

## Automated Review Classification Report

**Project:** Multilingual Mobile Health App Reviews (Arabic & English)

**Date:** October 8, 2025



# Few-Shot

## 1 Objective

This work aims to categorize a large user review dataset for a mobile health application automatically into Theme, Subtheme, and Sentiment classes using a **few-shot** machine learning approach.

This automated annotation helps in accelerating data preparation for further review and analysis of the user experience.

## 2 Data & Preprocessing

- **Dataset Size:** 120,561 reviews (Arabic + English).
- **Text Cleaning:** Removed noise, unified encoding, normalized spaces & emojis, and retained meaningful text. Missing Values Handling: All empty cells were filled with NA to maintain data integrity and avoid row loss (*additional step applied in this analysis*).
- **Sampling:**
  - A fixed **24,000-row stratified sample** was selected for model evaluation.
  - Few-shot training used **1,000 Arabic** and **500 English** labeled samples (when language information was available).



### Few-Shot Configuration

#### 🧠 Training Sizes

- Arabic (AR): **1000**
- English (EN): **500**
- Total sample for tests: **24,000**
- Random seed: **13**

#### ⚙️ Models & Features

- Subtheme: **LinearSVC** (TF-IDF 1–2 n-grams)
- Sentiment: **LogisticRegression** (class\_weight="balanced")
- Theme = mapped from Subtheme (fixed taxonomy)
- Fallback: ★ **Rating** → **Sentiment** if text not meaningful

#### 🏷️ Theme

Derived from Subtheme via canonical mapping

#### ✳️ Subtheme

LinearSVC + TF-IDF (1–2)

#### 😊 Sentiment

LogReg (balanced) + rating fallback



Writes predictions only to empty cells and preserves existing GT labels.

### 3 Classification Workflow

- **Theme Prediction:** Derived indirectly from Subtheme predictions using a fixed taxonomy mapping.
- **Subtheme Model:** Linear Support Vector Classifier (Linear SVC) with TF-IDF (1–2 n-grams).
- **sentiments model:** logistic regression with class balancing; fallback to star ratings where text is not important.
- **Few-shot Learning:** Utilized a small yet high-quality labeled dataset to develop efficient but strong models. Output Policy:
  - Only empty cells were filled; any existing ground-truth labels (GT) remained untouched.
  - Predictions were written back into a new Excel file with the suffix \_classification.

### 4 Evaluation — Stratified K-Fold Cross Validation (k=5)

A 5-fold **stratified** approach ensured balanced performance measurement across all categories.

Model Performance — Stratified K-Fold (k=5)

⚡ Category	🎯 Accuracy	🏆 Precision (Macro)	🔎 Recall (Macro)	▢ F1 (Macro)
👉 Theme	99.70%	93.99%	97.81%	95.61%
✳️ Subtheme	99.91%	95.30%	97.16%	96.03%
😊 Sentiment	99.93%	87.26%	89.97%	88.20%

Note: Since the dataset is highly imbalanced (≈approximately 80% positive sentiment), macro-averaged metrics (Precision/Recall/F1) were used to ensure a fair performance evaluation across classes. These results confirm that the models are not only accurate but also balanced

- ✓ These metrics indicate very high accuracy in both hierarchical topic classification (Theme/Subtheme) and sentiment prediction.



### Theme Predictions

Class	Count
User Experience & Sentiment	112423
Content & Services	4331
Technical Performance	3654
Security & Support	153

## 5 Key Results & Distribution

### Theme Predictions

- User Experience & Sentiment — 112,423
- Content & Services — 4,331
- Technical Performance — 3,654
- Security & Support — 153



### Subtheme Predictions

Class	Count
Overall Satisfaction	110661
Appointment Booking	4325
Crashes / Freezes	2521
Ease of Use	1484
Errors / Bugs	1052
Help & Guidance	195
Login / OTP	144
Navigation	67
General_Technical	56
Stability	16
Account Access Issues	9
appointment Booking	7
App Speed	6
Locations_Coverage	4
Teleconsultation	3
General_UX	3
Onboarding	2
Prescriptions	2
Connectivity / Network	2
Loading Time	1
Reports / Documents	1

### Subtheme Predictions (Top examples)

- Overall Satisfaction — 110,661
- Appointment Booking — 4,325
- Crashes / Freezes — 2,521
- Ease of Use — 1,484
- Errors / Bugs — 1,052



## Sentiment Predictions

Class	Count
Positive	83866
Negative	33984
Neutral	2711

## Sentiment Predictions

- Positive — 83,866
- Negative — 33,984
- Neutral — 2,711

## 6 Impact

- **Time saving:** Automated labeling replaced extensive manual work across 120k+ records.
- **Scalability:** The Approach can adapt to new data without full retraining.
- **Quality:** High macro-F1 (>95% for Theme/Subtheme) ensures reliable categorization for downstream analysis and visualization.

## Generated File:

Database\_Merged\_AR\_EN (sample24,000)\_classification.xlsx