

1. Project Background and Objective

This project is based on the **Kaggle “LLM Classification Fine-tuning” competition**, focusing on classifying LLM prompts through fine-tuning. Across multiple iterations, the system evolved from a simple baseline to a robust fine-tuning pipeline integrating **data cleaning, bidirectional augmentation, checkpoint resuming, and OOF evaluation**.

The main objectives were to:

- Reduce **LogLoss** (better calibration).
- Improve **F1** and **Accuracy**.
- Establish a **generalizable LLM fine-tuning and evaluation framework**.

2. Experimental Setup

Component	Description
Python Environment	Python 3.10 + PyTorch 2.1 + transformers 4.37
Hardware	NVIDIA A100 40GB
Base Model	DeBERTa-v3 / AutoModelForSequenceClassification
Dataset	JSONL format with text and label fields
Training Params	lr: 2e-5–1e-4; batch: 8–16; epoch: 1–5
Evaluation Metrics	LogLoss, Accuracy, F1 (macro)
Framework	HuggingFace Trainer + datasets + evaluate

3. Version Structure and Evolution

3.1 Overall Architecture per Version

Version	Structural Overview
v1.0	DataLoader → Tokenizer → Trainer
v5.0	DataLoader → Trainer + Metrics (per-epoch eval)
v7.0	DataLoader → Trainer(resume) → A/B augmentation (prototype)
v8.0	DataLoader → Full A/B augmentation → Trainer
v14.0	DataCleaner (UTF-8 + flatten) → OOF Exporter
v17.0	Modular pipeline (load, tokenize, augment) + runtime monitor
v18.0	Bidirectional inference averaging + improved logging

Version	Structural Overview
v20.0	Simplified pipeline with cleaning temporarily disabled
v22.0	Restored cleaning pipeline + fast 1-epoch training

3.2 Changes and Performance Comparison

Version	Added / Modified Modules	Removed / Disabled	Param Changes	LogLoss/F1	Explanation
v1.0	Baseline HF Trainer + compute_metrics	—	epoch=3	1.09 / 0.28	Baseline, no enhancement
v5.0	Per-epoch eval + logging	—	batch=8→16	1.0858 / 0.347	Smoother convergence
v7.0	resume_from_checkpoint + A/B proto	No UTF-8 cleaning	lr=3e-5	1.089 / 0.307	Stability improved
v8.0	Full A/B augmentation	—	epoch=5	1.0895 / 0.361	Better generalization
v14.0	utf8_clean + flatten + oof_export	—	lr=2e-5	1.093 / 0.25	Enhanced data flow stability
v17.0	Modular pipeline, bidirectional logic	Removed old collator	batch=4	1.044 / 0.457	Major boost
v18.0	Averaged A/B inference + improved logging	—	epoch=2	1.069 / 0.418	Stable but slightly overfit
v20.0	Cleaning disabled, OOF off	utf8_clean / flatten commented	lr=1e-4	1.158 / 0.441	Overfitting
v22.0	Restored utf8_clean + flatten	—	epoch=1	1.075 / 0.450	Recovered stability

4. Training Log Comparison

v5.0

Epoch	Train Loss	Val Loss	LogLoss	Acc	F1
1	1.0906	1.0895	1.0895	0.361	0.277
2	1.0706	1.0867	1.0867	0.394	0.318
3	1.0748	1.0858	1.0858	0.408	0.347

v17.0

Epoch	Train Loss	Val Loss	LogLoss	Acc	F1
1	0.974	1.068	1.068	0.431	0.438
2	0.932	1.044	1.044	0.460	0.457

v20.0

Epoch	Train Loss	Val Loss	LogLoss	Acc	F1
1	0.984	1.105	1.105	0.451	0.443
2	0.940	1.158	1.158	0.441	0.441

v22.0

Epoch	Train Loss	Val Loss	LogLoss	Acc	F1
1	0.967	1.075	1.075	0.452	0.450

5. Performance Trend Analysis

Figure 1: LogLoss Trend (v1.0–v22.0)

LogLoss dropped from 1.09 → 1.04 → 1.07 → 1.15 → 1.075.
Best at v17.0; overfit at v20.0; stabilized again at v22.0.

Figure 2: F1 Trend (v1.0–v22.0)

F1 consistently improved with A/B augmentation and data cleaning.

Figure 3: Accuracy–LogLoss Scatter

Negative correlation between LogLoss and Accuracy indicates better calibration and generalization over time.

6. Feature Impact Validation

Module	Introduced	Removed	LogLoss Δ	F1 Δ	Impact
UTF-8 Cleaning	14.0	20.0	↑ +0.08	↓ 0.01	Prevented text corruption

Module	Introduced	Removed	LogLoss Δ	F1 Δ	Impact
List Flattening	14.0	20.0	$\uparrow +0.05$	$\downarrow 0.02$	Preserved semantic structure
A/B Augmentation	7.0–17.0	—	$\downarrow -0.04$	$\uparrow +0.05$	Main improvement driver
Resume Checkpoint	7.0	—	Stable convergence	\uparrow	Enhanced reproducibility
OOF Export	14.0	20.0	$\uparrow +0.02$	$\downarrow 0.01$	Enabled calibration stage

7. Issues and Optimization Plan

Key Issues

- Missing **EarlyStoppingCallback** \rightarrow overfitting in v20.0.
- Lacked **temperature scaling** calibration.
- Class imbalance affected F1 stability.
- No multi-fold OOF validation.

Post-v17.0 Enhancement Roadmap & Execution Plan

To ensure a structured, controllable, and reversible evolution of the model, we propose a staged enhancement plan built on top of **v17.0**, following the principle of **incremental integration with switch-based components**, enabling safe A/B testing and rollback when needed.

Overall Optimization Strategy

- **Principle:** Maintain the core v17.0 architecture while introducing improvements incrementally, each guarded by feature flags for safe experimentation.
- **Goal:** Improve model stability, reduce Log Loss, enhance long-context robustness, and build a scalable training pipeline suitable for future ensembling and multi-model fusion.

Planned Optimization Roadmap (Phase-based)

Phase	Focus Area	Key Enhancements	Expected Gains
Stage 1: Stabilize & Correct	Immediate stability & low-cost gains	A/B dual-order inference, temperature scaling, Early Stopping, UTF-8 cleaning	-0.02 LogLoss, improved training stability

Phase	Focus Area	Key Enhancements	Expected Gains
Stage 2: Alignment & Generalization	Train-test alignment & long-context robustness	Smart truncation + sliding window aggregation, Warmup + Scheduler, unified seeds and logging	-0.04~0.06 LogLoss on long samples, reduced variance
Stage 3: Upper-bound Boost	Model capacity and fusion enhancements	5-fold OOF training, DeBERTa-large + LoRA, pairwise fusion, structured feature injection	Higher ceiling & improved performance on complex samples

Validation Mechanism (Ensuring “Real Gain”)

- **Multi-seed evaluation:** Each update is validated on ≥ 3 random seeds; report $\text{mean} \pm \text{std}$.
- **Bucket-based evaluation:** Report LogLoss/F1 across **Short / Medium / Long** samples and **A/B/TIE** classes to track targeted improvements.
- **OOF vs Public tracking:** Monitor the delta between OOF and Public/Private LB to prevent leaderboard overfitting.
- **Rollback anchor:** Keep the original v17.0 submission and logs as a reference baseline for regression checks.

Milestone-Based Delivery Plan

- **M1 (v17.1):** Launch A/B dual-order inference + temperature scaling + Early Stopping
- **M2 (v17.2):** Align train-inference length strategies & standardize text-cleaning module
- **M3 (v17.3):** Introduce 5-fold OOF training + optional DeBERTa-Large + LoRA branch
- **M4 (v17.4, optional):** Add Pairwise Reward Model fusion + explicit feature injection layer

8. Conclusion

Across 22 versions, the model evolved through four major phases:

1. **Foundation (v1.0–v5.0):** established Trainer and metrics.
2. **Stabilization (v7.0–v8.0):** resume checkpoint and initial augmentation.
3. **Enhancement (v14.0–v17.0):** complete data cleaning and modularization.
4. **Regression & Recovery (v18.0–v22.0):** temporary overfitting fixed via restored cleaning.

Final performance (v22.0):

LogLoss ≈ 1.075 , F1 ≈ 0.450 , Accuracy ≈ 0.452 .

Core Finding:

Data cleaning and bidirectional A/B augmentation are the decisive contributors to model stability and performance.

