

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	↑ +0.05	↓ 0.02	Preserved semantic structure
A/B Augmentation	7.0–17.0	—	↓ -0.04	↑ +0.05	Main improvement driver
Resume Checkpoint	7.0	—	Stable convergence	↑	Enhanced reproducibility
OOF Export	14.0	20.0	↑ +0.02	↓ 0.01	Enabled calibration stage

7. Issues and Optimization Plan

Key Issues

- Missing **EarlyStoppingCallback** → 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 mean \pm 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.

9. Appendix

9.1 Full Metrics per Version

Version	LogLoss	Accuracy	F1
1.0	1.09	0.36	0.28
5.0	1.0858	0.408	0.347
7.0	1.089	0.384	0.307
8.0	1.0895	0.369	0.361
14.0	1.093	0.370	0.250
17.0	1.044	0.460	0.457
18.0	1.069	0.421	0.418
20.0	1.158	0.441	0.441
22.0	1.075	0.452	0.450

9.2 Feature Matrix

Feature	1.0	5.0	7.0	8.0	14.0	17.0	18.0	20.0	22.0
Resume Training									
A/B Augmentation									
UTF-8 Cleaning									
List Flattening									
OOF Export									
Early Stopping									
Temperature Calibration									