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CoorLog: Efficient-Generalizable Log Anomaly Detection via Adaptive Coordinator in Software Evolution

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1. Background & Motivation

Log anomaly detection (AD) is crucial for ensuring system reliability.

- **Problem**: Frequent software updates change log structures and patterns.

Motivation

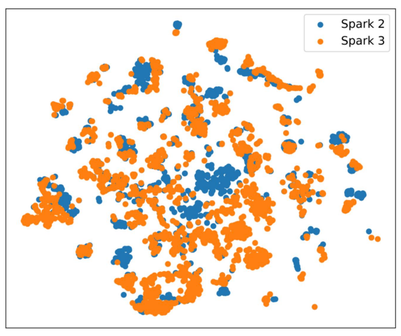
- **SM :** Cost 👍 Genrealization 👎
- **LLM:** Cost 👎 Genrealization 👍

Case 1 Log Entry Evolution

Spark 2	Started reading broadcast variable <*>
Spark 3	Started reading broadcast variable <*> with <*> pieces(estimated total size <*> MiB)

Case 2 Log Sequences Evolution

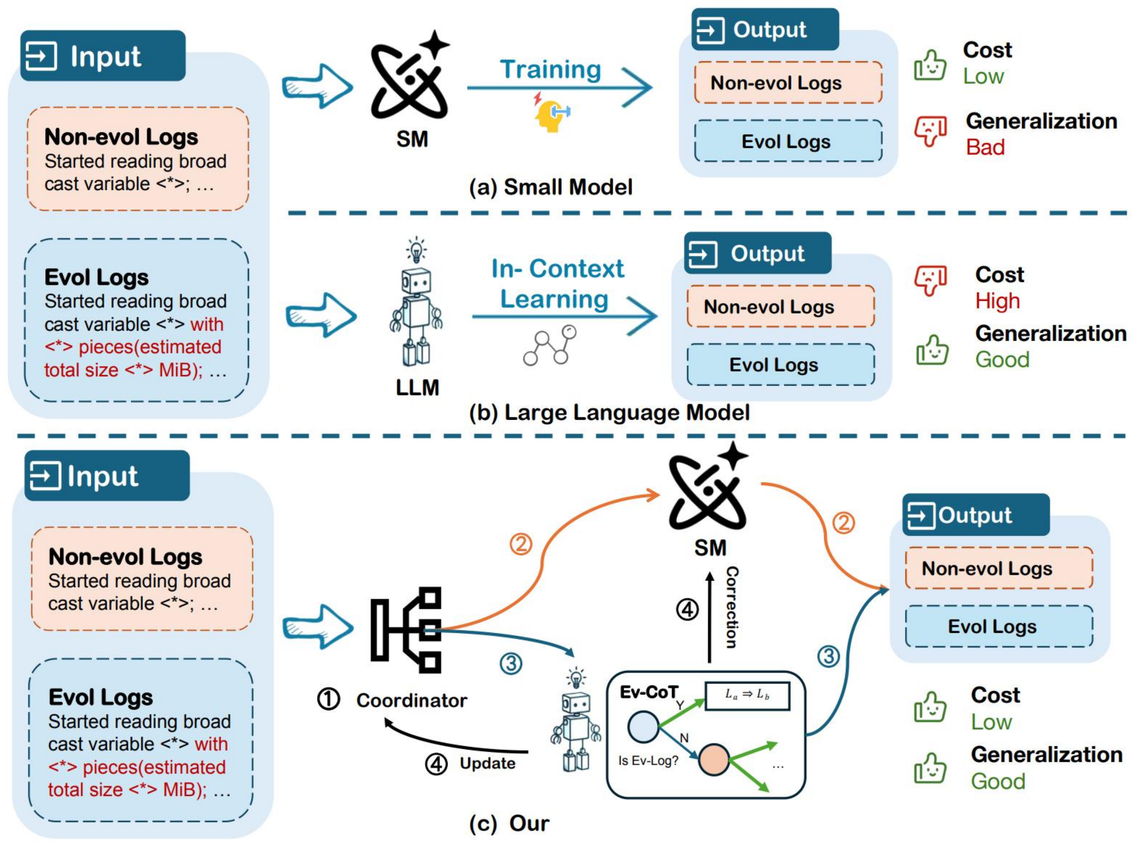
Spark 2 E1 → E3	E1: Connecting to driver: <*> E2: Successfully registered with driver E3: Resources for <*>:
Spark 3 E1 → E2 → E3	



(a) Spark

Proposed Solution

- Collaboration between LLM and SM to balance efficiency and generalization via the Adaptive Coordinator.



2. Method

Stage 1: Log Identification via the Coordinator

Use an **AutoEncoder** to distinguish between known and new log pattern.

Stage 2: Non-Evolved Log Detection via SM

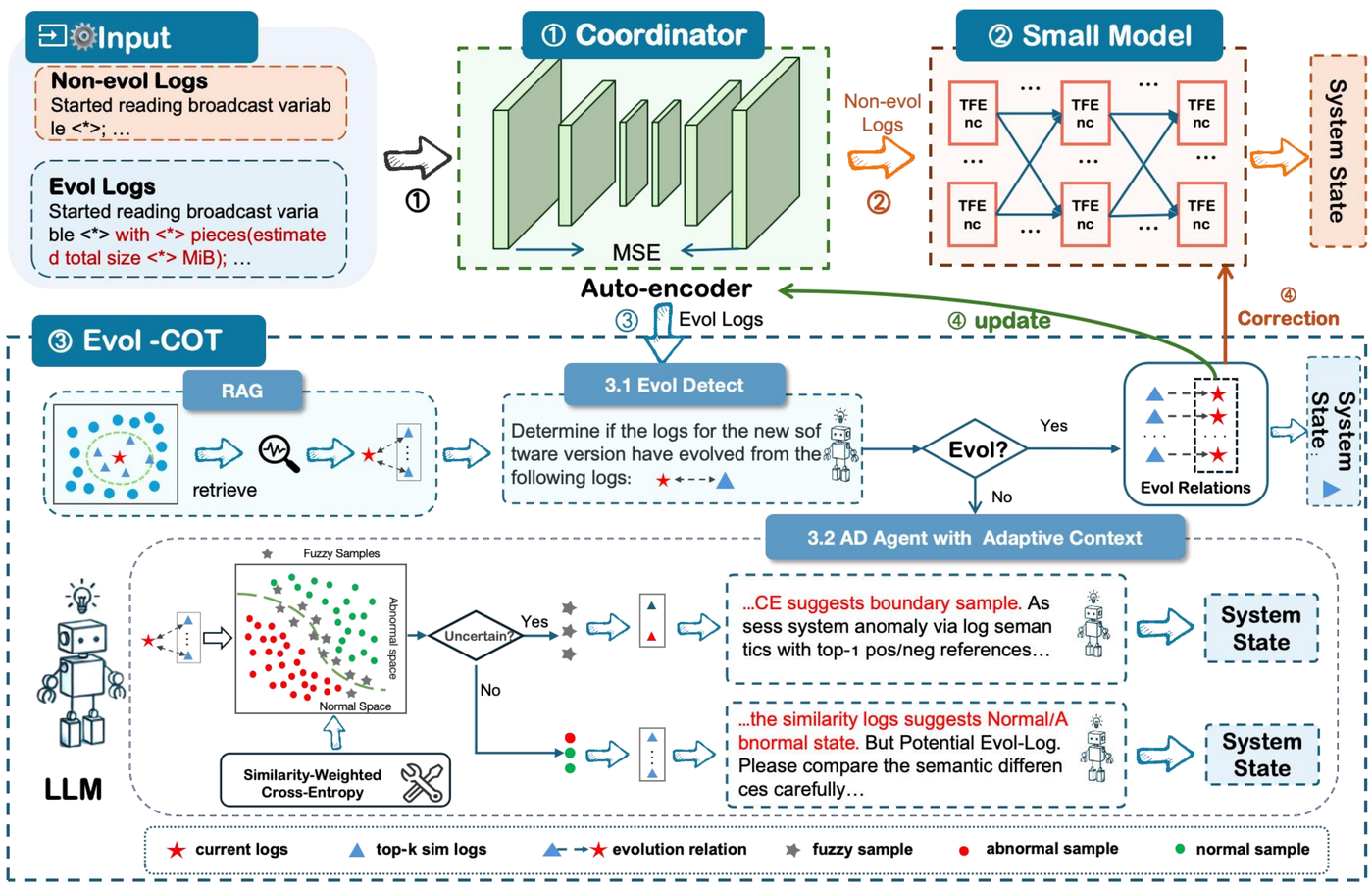
Use a **BERT-based** Network to handle in-distribution logs.

Stage 3: Evolved Log Detection via LLM

Design the **Evol-CoT** framework for fine-grained inference on evolved logs.

Stage 4: Adaptive Evolution Mechanism

Use **AEM** to avoid redundant inference for the same samples in LLM.



3. Experiment and Result

Key Conclusion

- **The coordinator can identify concept drift logs.**
- **Evol-CoT provides interpretability.**
- **CoorLog achieves a high F1-score with low cost, making it suitable for real-world deployment.**

LogEvol-Hadoop

Method	Intra-version: Hadoop 2 → Hadoop 2 Pr	Re	F1	Intra-version: Hadoop 3 → Hadoop 3 Pr	Re	F1	Inter-version: Hadoop 2 → Hadoop 3 Pr	Re	F1	Inter-version: Hadoop 3 → Hadoop 2 Pr	Re	F1
LogSed	0.910	**0.995**	0.951	0.925	0.986	0.955	0.371	0.988	0.540	0.390	0.993	0.560
DeepLog	0.913	0.985	0.947	0.926	**1.000**	0.961	0.386	**0.999**	0.556	0.410	0.971	0.576
LogAnomaly	0.926	0.994	0.958	0.939	0.988	0.963	0.389	0.998	0.560	0.407	**0.995**	0.578
BERT	0.928	0.731	0.817	0.959	0.837	0.894	0.865	0.706	0.778	0.952	0.763	0.847
LogRobust	0.935	0.981	0.957	0.948	0.983	0.965	0.782	0.824	0.803	0.813	0.846	0.829
LogBERT	0.941	0.977	0.959	0.953	0.987	0.970	0.875	0.852	0.863	0.898	0.871	0.884
LogOnline	0.948	0.984	0.966	0.963	0.989	0.976	0.893	0.895	0.894	0.913	0.908	0.911
LLMeLog	0.952	0.967	0.959	0.963	0.975	0.969	0.912	0.923	0.917	0.928	0.934	0.931
EvLog	0.945	0.982	0.963	0.952	0.988	0.970	0.770	0.941	0.847	0.857	0.913	0.884
Our	**0.993**	0.968	**0.980**	**0.997**	0.982	**0.990**	**0.946**	0.983	**0.964**	**0.994**	0.957	**0.975**

LogEvol-Spark

Method	Intra-version: Spark 2 → Spark 2 Pr	Re	F1	Intra-version: Spark 3 → Spark 3 Pr	Re	F1	Inter-version: Spark 2 → Spark 3 Pr	Re	F1	Inter-version: Spark 3 → Spark 2 Pr	Re	F1
LogSed	0.842	0.914	0.877	0.907	0.923	0.915	0.013	0.917	0.026	0.010	0.914	0.020
DeepLog	0.862	**0.952**	0.905	0.858	**0.976**	0.914	0.017	**0.947**	0.032	0.014	0.909	0.026
LogAnomaly	0.931	0.939	0.935	0.898	0.947	0.922	0.020	0.923	0.038	0.017	**0.948**	0.034
BERT	0.943	0.750	0.835	**1.000**	0.684	0.812	0.550	0.696	0.615	**1.000**	0.568	0.715
LogRobust	0.949	0.837	0.889	0.974	0.857	0.912	0.732	0.753	0.742	0.934	0.783	0.851
LogBERT	0.948	0.875	0.909	0.973	0.886	0.927	0.805	0.813	0.809	0.949	0.822	0.881
LogOnline	0.954	0.904	0.928	0.986	0.899	0.940	0.843	0.864	0.853	0.957	0.875	0.914
LLMeLog	0.959	0.920	0.938	0.950	0.857	0.900	0.840	0.799	0.828	0.762	0.934	0.851
EvLog	0.970	**0.974**	**0.972**	0.944	0.888	0.915	0.922	0.700	0.795	0.920	0.812	0.863
Our	**0.976**	0.932	0.954	0.979	0.918	**0.947**	**0.904**	0.855	**0.879**	0.968	0.904	**0.935**

Table I Comparison of different strategies

Logs	Method	Spark 2 → 3 Pr	Re	F1	Hadoop 2 → 3 Pr	Re	F1
Non-evol	Our_{SM}	0.950	0.901	0.925	0.985	0.979	0.982
	+AEM	0.935	0.925	0.930	0.955	0.943	0.949
Evol	Our_{SM}	0.468	0.417	0.441	0.375	0.402	0.387
	+AEM	0.640	0.610	0.625	0.537	0.489	0.512
	Vanilla	0.785	0.209	0.330	0.705	0.518	0.595
	Evol-CoT	0.829	0.913	0.870	0.944	0.885	0.914
ALL	Our_{SM}	0.698	0.848	0.766	0.893	0.861	0.876
	+AEM	0.721	0.867	0.794	0.895	0.872	0.883
	Our	0.904	0.855	0.879	0.946	0.983	0.964

Table II Ablation study on different log categories

	#Relations	Accuracy	AEM Time (ms)
Spark 2 → 3	288	0.993	0.18
Hadoop 2 → 3	541	0.986	0.02

Table III Statistics of evolution relations

Metric	Method	Spark 2 → 3	Hadoop 2 → 3
Calls	w/o coord	4,246	34,302
	Our	289	2,174
Token (k)	w/o coord	9,048.85	21,347.73
	Our	1,557.30	2,480.65
Time (ms)	LogRobust	0.96	0.43
	LogAnomaly	1.69	0.83
	Our_{SM}	0.23	0.04
	w/o coord	810.10	724.61
	Our	78.40	54.30

Table IV Comparison of efficiency and cost

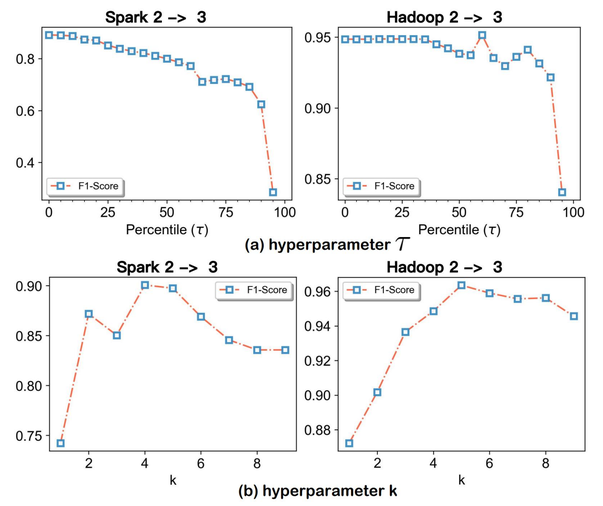


Fig. 6: Ablation study on hyperparameter (a) τ and (b) k.