

ARES: Anomaly Recognition Model For Edge Streams

Online Appendix

Model	DARPA		UNSW-NB15		ISCX2012		CIC-IDS2017		CTU-13					
									Scenario 1		Scenario 10		Scenario 13	
	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP
MIDAS-F	0.961 \pm 0.005	0.979 \pm 0.003	0.793 \pm 0.003	0.337 \pm 0.004	0.974 \pm 0.004	0.287 \pm 0.031	0.997 \pm 0.000	0.969 \pm 0.014	0.835 \pm 0.037	0.077 \pm 0.014	0.924 \pm 0.020	0.530 \pm 0.078	0.934 \pm 0.050	0.135 \pm 0.052
AnoEdge-L	0.910 \pm 0.007	0.955 \pm 0.003	0.750 \pm 0.021	0.443 \pm 0.033	0.869 \pm 0.038	0.097 \pm 0.027	0.954 \pm 0.006	0.486 \pm 0.028	0.479 \pm 0.007	0.018 \pm 0.001	0.692 \pm 0.006	0.181 \pm 0.010	0.293 \pm 0.009	0.010 \pm 0.000
SLADE-H	0.900 \pm 0.093	0.912 \pm 0.091	0.911 \pm 0.118	0.757 \pm 0.191	0.999 \pm 0.000	0.985 \pm 0.022	0.731 \pm 0.226	0.215 \pm 0.178	0.564 \pm 0.077	0.018 \pm 0.005	0.399 \pm 0.159	0.099 \pm 0.020	0.564 \pm 0.077	0.018 \pm 0.005
ARES-Static	0.991 \pm 0.005	0.994 \pm 0.003	0.984 \pm 0.003	0.885 \pm 0.016	0.999 \pm 0.000	0.945 \pm 0.010	0.984 \pm 0.001	0.811 \pm 0.017	0.970 \pm 0.003	0.296 \pm 0.030	0.945 \pm 0.024	0.701 \pm 0.088	0.928 \pm 0.122	0.282 \pm 0.119
ARES-Dynamic	0.989 \pm 0.007	0.994 \pm 0.003	0.877 \pm 0.009	0.500 \pm 0.030	0.999 \pm 0.000	0.960 \pm 0.018	0.984 \pm 0.001	0.816 \pm 0.013	0.947 \pm 0.027	0.209 \pm 0.073	0.924 \pm 0.022	0.673 \pm 0.091	0.851 \pm 0.042	0.098 \pm 0.040

Table 1: Comparison of ARES against competing methods by evaluating performance on subsampled versions of the test set for each dataset.

Model	DARPA		UNSW-NB15		ISCX2012		CIC-IDS2017		CTU-13					
									Scenario 1		Scenario 10		Scenario 13	
	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP	ROC-AUC	AP
GNN-RecErr	0.304 \pm 0.000	0.034 \pm 0.000	0.629 \pm 0.000	0.247 \pm 0.000	0.559 \pm 0.000	0.059 \pm 0.000	0.419 \pm 0.000	0.126 \pm 0.000	0.394 \pm 0.000	0.022 \pm 0.000	0.222 \pm 0.000	0.041 \pm 0.000	0.487 \pm 0.000	0.023 \pm 0.000
GCN+HST	0.637 \pm 0.101	0.806 \pm 0.039	0.986 \pm 0.001	0.894 \pm 0.004	0.969 \pm 0.009	0.347 \pm 0.085	0.881 \pm 0.061	0.352 \pm 0.096	0.707 \pm 0.173	0.066 \pm 0.032	0.348 \pm 0.071	0.138 \pm 0.010	0.817 \pm 0.059	0.068 \pm 0.019
GAT+HST	0.965 \pm 0.004	0.975 \pm 0.003	0.688 \pm 0.296	0.521 \pm 0.289	0.839 \pm 0.072	0.138 \pm 0.058	0.974 \pm 0.024	0.734 \pm 0.140	0.654 \pm 0.009	0.034 \pm 0.005	0.914 \pm 0.076	0.592 \pm 0.194	0.791 \pm 0.002	0.037 \pm 0.001
GraphSage+RRCF	0.947 \pm 0.041	0.955 \pm 0.035	0.811 \pm 0.059	0.538 \pm 0.069	0.995 \pm 0.001	0.679 \pm 0.011	0.616 \pm 0.097	0.347 \pm 0.059	0.804 \pm 0.064	0.072 \pm 0.025	0.728 \pm 0.041	0.226 \pm 0.036	0.506 \pm 0.105	0.025 \pm 0.005
ARES	0.985 \pm 0.008	0.991 \pm 0.005	0.985 \pm 0.002	0.892 \pm 0.014	0.999 \pm 0.000	0.942 \pm 0.007	0.984 \pm 0.001	0.806 \pm 0.025	0.969 \pm 0.004	0.291 \pm 0.014	0.956 \pm 0.019	0.712 \pm 0.086	0.948 \pm 0.049	0.249 \pm 0.091

Table 2: Ablation study results on multiple datasets, highlighting the effectiveness of combining GraphSAGE embeddings with Half-Space Trees.

Dataset	ARES	RRCF
DARPA	7.8	33.9
UNSW-NB15	5.4	10.7
ISCX2012	14.1	28.3
CIC-IDS2017	51.4	457.6
CTU-13 Scenario 1	59.5	260.2
CTU-13 Scenario 10	44.9	85.4
CTU-13 Scenario 13	108.3	203.5

Table 3: Average inference time (in seconds) across multiple datasets for ARES (HST-based) and RRCF.