

ARES: Anomaly Recognition Model For Edge Streams

Online Appendix

Model	DARPA				UNSW-NB15				ISCX2012				CIC-IDS2017				CTU-13						
	ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		
MIDAS-F	0.961 ± 0.005	0.979 ± 0.003	0.793 ± 0.003	0.337 ± 0.004	0.974 ± 0.004	0.287 ± 0.031	0.997 ± 0.000	0.969 ± 0.014	0.835 ± 0.037	0.077 ± 0.014	0.924 ± 0.020	0.530 ± 0.078	0.934 ± 0.050	0.135 ± 0.052									
AnoEdge-L	0.910 ± 0.007	0.955 ± 0.003	0.750 ± 0.021	0.443 ± 0.033	0.869 ± 0.038	0.097 ± 0.027	0.954 ± 0.006	0.486 ± 0.028	0.479 ± 0.007	0.018 ± 0.001	0.692 ± 0.006	0.181 ± 0.010	0.293 ± 0.009	0.010 ± 0.000									
SLADE-H	0.900 ± 0.093	0.912 ± 0.091	0.911 ± 0.118	0.757 ± 0.191	0.999 ± 0.000	0.985 ± 0.022	0.731 ± 0.226	0.215 ± 0.178	0.564 ± 0.077	0.018 ± 0.005	0.399 ± 0.159	0.099 ± 0.020	0.564 ± 0.077	0.018 ± 0.005									
ARES-Static	0.991 ± 0.005	0.994 ± 0.003	0.984 ± 0.003	0.885 ± 0.016	0.999 ± 0.000	0.945 ± 0.010	0.984 ± 0.001	0.811 ± 0.017	0.970 ± 0.003	0.296 ± 0.030	0.945 ± 0.024	0.701 ± 0.088	0.928 ± 0.122	0.282 ± 0.119									
ARES-Dynamic	0.989 ± 0.007	0.994 ± 0.003	0.877 ± 0.009	0.500 ± 0.030	0.999 ± 0.000	0.960 ± 0.018	0.984 ± 0.001	0.816 ± 0.013	0.947 ± 0.027	0.209 ± 0.073	0.924 ± 0.022	0.673 ± 0.091	0.851 ± 0.042	0.098 ± 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							
	ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		AP		ROC-AUC		AP	
GNN-RecErr	0.304 ± 0.000	0.034 ± 0.000	0.629 ± 0.000	0.247 ± 0.000	0.559 ± 0.000	0.059 ± 0.000	0.419 ± 0.000	0.126 ± 0.000	0.394 ± 0.000	0.022 ± 0.000	0.222 ± 0.000	0.041 ± 0.000	0.487 ± 0.000	0.023 ± 0.000										
GCN+HST	0.637 ± 0.101	0.806 ± 0.039	0.986 ± 0.001	0.894 ± 0.004	0.969 ± 0.009	0.347 ± 0.085	0.881 ± 0.061	0.352 ± 0.096	0.707 ± 0.173	0.066 ± 0.032	0.348 ± 0.071	0.138 ± 0.010	0.817 ± 0.059	0.068 ± 0.019										
GAT+HST	0.965 ± 0.004	0.975 ± 0.003	0.688 ± 0.296	0.521 ± 0.289	0.839 ± 0.072	0.138 ± 0.058	0.974 ± 0.024	0.734 ± 0.140	0.654 ± 0.009	0.034 ± 0.005	0.914 ± 0.076	0.592 ± 0.194	0.791 ± 0.002	0.037 ± 0.001										
GraphSage+RRCF	0.947 ± 0.041	0.955 ± 0.035	0.811 ± 0.059	0.538 ± 0.061	0.995 ± 0.001	0.679 ± 0.011	0.616 ± 0.097	0.347 ± 0.059	0.804 ± 0.064	0.072 ± 0.025	0.728 ± 0.041	0.226 ± 0.036	0.506 ± 0.105	0.025 ± 0.005										
ARES	0.985 ± 0.008	0.991 ± 0.005	0.985 ± 0.002	0.892 ± 0.014	0.999 ± 0.000	0.942 ± 0.007	0.984 ± 0.001	0.806 ± 0.025	0.969 ± 0.004	0.291 ± 0.014	0.956 ± 0.019	0.712 ± 0.086	0.948 ± 0.049	0.249 ± 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.