## Supplementary Material for "OML-AD: Online Machine Learning for Anomaly Detection in Time Series Data"

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## 1 Supplementary Figures

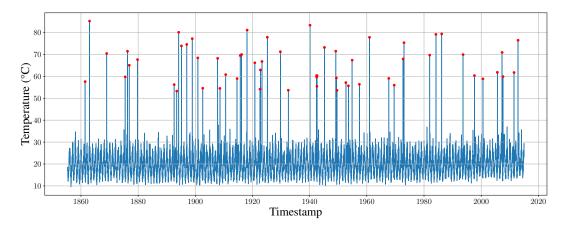
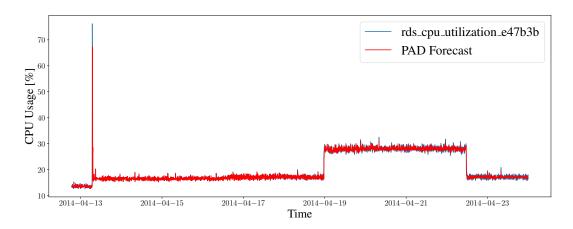
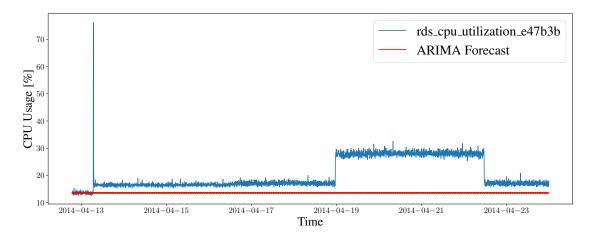


Fig. 1. Weekly Temperature Data with Synthesized Anomalies



 ${\bf Fig.\,2.}$  OML-AD Forecast on CPU Utilization Data



 ${\bf Fig.\,3.}$  SARIMA Forecast on CPU Utilization Data without Retraining

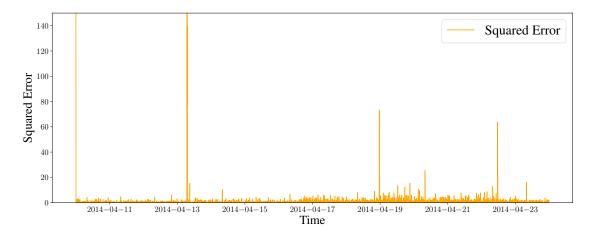


Fig. 4. Error of OML-AD Forecast on CPU Utilization Data

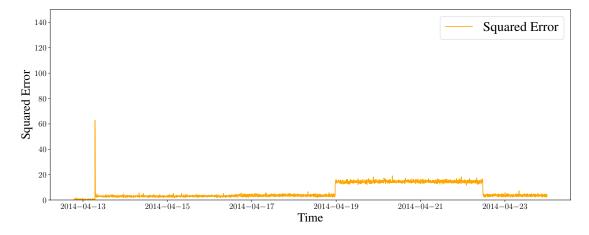


Fig. 5. Error of SARIMA Forecast on CPU Utilization Data

## 2 Supplementary Tables

Table 1. Forecasting and detection performance on weather data with synthesized anomalies

City	Algorithm		MAE	MSE	F1	AUC ROC
Sydney	OML-AD		2.7504	8.0843	0.9503	0.9879
	SARIMA	No Retraining	6.3630	69.0261	0.1320	0.9765
		Scheduled Retraining	2.5258	21.9888	0.6170	0.9861
		Dynamic Retraining	2.4962	20.9147	0.8862	0.9968
	Prophet	No Retraining	16.3098	387.5487	0.0398	0.8558
		Scheduled Retraining	6.5243	68.9949	0.7420	0.9651
		Dynamic Retraining	2.5856	23.6932	0.8025	0.9677
	OML-AD		2.7637	7.9064	0.9747	0.9998
ne	SARIMA	No Retraining	6.0970	66.6365	0.1370	0.9584
ur		Scheduled Retraining	2.5819	22.5574	0.5987	0.9957
Melbourne		Dynamic Retraining	2.4129	21.1346	0.9014	0.9989
$\mathbb{X}$	Prophet	No Retraining	17.0762	404.8067	0.0402	0.8425
		Scheduled Retraining	6.6228	69.2407	0.7230	0.9975
		Dynamic Retraining	2.6378	23.7981	0.8318	0.9987
Robe	OML-AD		2.6104	7.5719	0.9719	0.9988
	SARIMA	No Retraining	6.4173	68.7132	0.1372	0.9490
		Scheduled Retraining	2.5203	23.6533	0.5857	0.9942
		Dynamic Retraining	2.5432	20.1169	0.8599	0.9939
	Prophet	No Retraining	18.0550	397.8834	0.0389	0.8043
		Scheduled Retraining	6.6134	66.2162	0.7011	0.9454
		Dynamic Retraining	2.4831	24.0231	0.8046	0.9621

Table 2. Forecasting and detection performance on CPU utility data with real anomalies

Algorithm	l	MAE	MSE	F1	AUC ROC
OML-AD		0.7525	2.4217	0.4444	0.9992
SARIMA	No Retraining Scheduled Retraining Dynamic Retraining	6.7164 4.2726 1.2050	39.9807		0.8438 0.8420 0.9906
Prophet	No Retraining Scheduled Retraining Dynamic Retraining		99.8151 29.1455 470.6927	0.5000	0.8438 0.8686 0.7545

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Table 3. Time and resource consumption on weather data with synthesized anomalies

City	Algorithm	ı	Mean Time [ms]	Std [ms]	CPU [%]	RAM [%]
Sydney	OML-AD		628.83	364.26	3.95	22.09
		No Retraining	58913.33	774.78	15.05	29.52
	SARIMA	Scheduled Retraining	164313.09	2098.46	15.83	29.10
		Dynamic Retraining	344827.70	4911.51	15.23	30.57
	Prophet	No Retraining	2078.27	459.57	4.20	33.21
		Scheduled Retraining	6482.84	2669.68	12.91	29.42
		Dynamic Retraining	12132.69	1178.52	11.95	29.21
	OML-AD		660.78	352.42	4.16	23.00
ne	SARIMA	No Retraining	57674.96	741.55	14.78	30.92
our		Scheduled Retraining	164430.49	2013.63	15.79	30.01
Melbourne		Dynamic Retraining	340427.43	4686.40	15.32	29.71
Ĭ	Prophet	No Retraining	2173.90	460.67	4.22	31.76
		Scheduled Retraining	6445.34	2715.17	13.33	30.41
		Dynamic Retraining	12274.80	1149.53	11.78	29.84
Robe	OML-AD		699.30	353.24	4.16	21.86
	SARIMA	No Retraining	60707.34	781.29	15.52	32.69
		Scheduled Retraining	155240.83	1941.19	15.93	28.45
		Dynamic Retraining	342832.87	4883.07	14.80	29.97
	Prophet	No Retraining	2171.43	445.04	4.06	31.07
		Scheduled Retraining	6506.64	2823.17	13.42	29.28
		Dynamic Retraining	11824.89	1115.07	11.36	30.04

Table 4. Time and resource consumption on CPU utility data with real anomalies

Algorithm		Mean Time [ms]	Std [ms]	CPU [%] I	RAM [%]
OML-AD		154.96	7.04	2.82	31.38
SARIMA	No Retraining Scheduled Retraining Dynamic Retraining		1128.20 6034.02 3293.89	6.13 9.71 9.99	48.11 41.56 39.81
Prophet	No Retraining Scheduled Retraining Dynamic Retraining	592.08 2194.62 4442.64	33.82 579.22 260.73	2.39 9.04 7.09	42.04 41.18 41.43