

# Random Similarity Isolation Forests: Supplementary Results

## S1 Datasets

The used datasets were taken from outlier detection benchmarks and generators for different types of data:

- numerical and categorical data: <https://github.com/Minqi824/ADBench>
- time series: <https://outlier-detection.github.io/utsd>
- sequences of sets: <https://gingerbread.shinyapps.io/SequencesOfSetsGenerator/>
- images, text, and graphs: <https://github.com/GuansongPang/ADRepository-Anomaly-detection-datasets>

The exact versions of the datasets can be found in the code repository accompanying this paper. When looking for benchmarks, we favored those in which the examples marked as outliers constituted 5% or fewer examples in the dataset.

**Sensitivity analysis.** For the analysis of the hyperparameters of RSIF, we used datasets that were independent of those used for the experimental comparison with other methods. More precisely, we used 10 datasets: 4 numerical (`cardio`, `lymphography`, `optdigits`, `speech`), 1 categorical (`ad`), 3 graph (`cox2`, `bzr`, `dhfr`), 1 text embedding (`agnews`), and 1 time-series (`twoleadecg`).

**Scalar data.** For tests with scalar data, we used 10 popular benchmark datasets, including 5 numerical (`glass`, `musk`, `satimage`, `vowels`, `wbc`) and 5 categorical (`aid`, `apascal`, `cmc`, `reuters`, `solarflare`). The datasets were chosen for their variety in terms of the number of examples and features (Table S1).

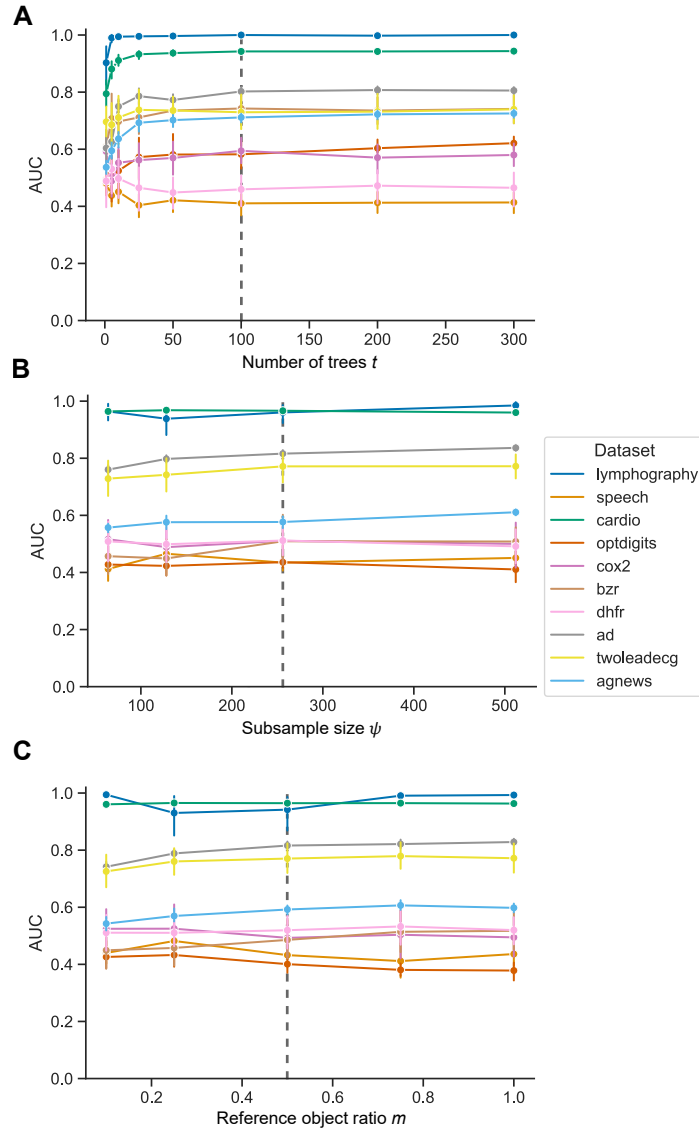
**Complex data.** For experiments with complex objects, we used 5 graph (`aids`, `dd`, `enzymes`, `nci1`, `proteins`), 2 time series (`earthquakes`, `aibo`), 3 text (`amazon`, `imdb`, `yelp`), and 3 image (`cifar`, `fashionmnist`, `svhn`) datasets. For the text and image datasets we used embeddings of pretrained RoBERTa<sup>1</sup> and ViT<sup>2</sup> models, respectively.

**Mixed data.** For mixed-type data, we used 3 sequences of sets (`item`, `length`, `order`) and 3 multiomics (`ovarian`, `her2`, `rosmap`) datasets. Sequences of sets, by nature, can be treated as sets, as sequences, or as combinations of the two representations. The multiomics datasets consist of the results of different genetic measurements, represented as distributions (lengths of variants) or numbers (gene expression).

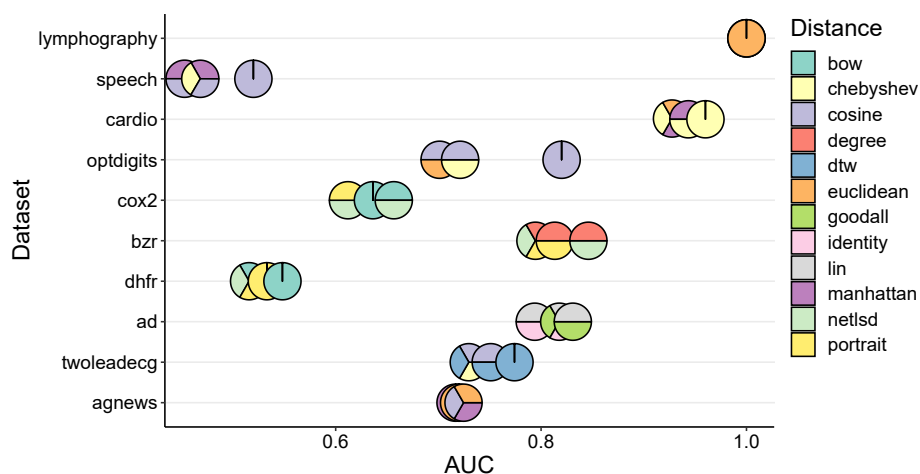
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<sup>1</sup>Liu, Y., Ott, M., Goyal, N., Du, J., Joshi, M., Chen, D., Levy, O., Lewis, M., Zettlemoyer, L., Stoyanov, V.: Roberta: A robustly optimized BERT pretraining approach. **CoRR abs/1907.11692**, 1–13 (2019)

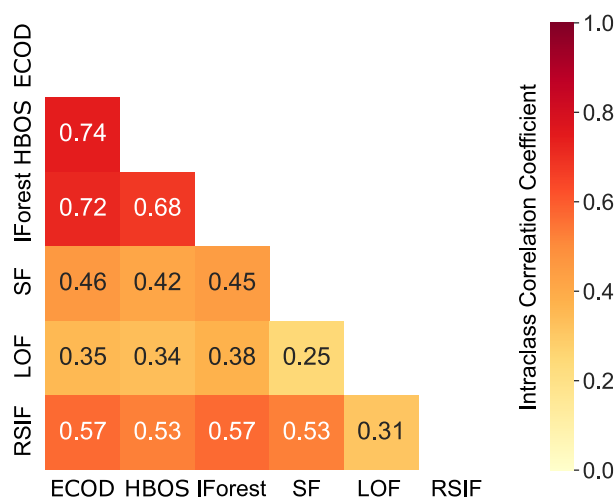
<sup>2</sup>Dosovitskiy, A., et al.: An image is worth 16x16 words: Transformers for image recognition at scale. **CoRR abs/2010.11929**, 1–22 (2020)



**Fig. S1.** Hyperparameter sensitivity analysis. Effect of (A) the number of trees  $t$ , (B) subsample size  $\psi$ , and (C) pool of reference objects on RSIF's predictive performance (ROC AUC). Bars represent 95% confidence intervals. The dashed gray lines show the selected defaults.



**Fig. S2.** Top 3 distance measure combinations for each of the sensitivity test datasets. Each set of measures used by RSIF is represented by a circle. If the set of distances consists of more than one measure, the circle is divided into multiple colored pieces, with colors defining the measures.



**Fig. S3.** Pairwise average intraclass correlation coefficients (ICC) based on outlier scores for the holdout test sets.

**Table S1.** AUC performance of RSIF and five competitive methods. The best results on each dataset are highlighted in bold, and the second best are underlined.

Dataset	Type	Features	#Ex.	#Feat.	%Outlier	AUC					
						IF	LOF	HBOSECOD	SF	RSIF	
glass	scalar	numeric	214	7	4.21	0.72	0.74	0.76	0.60	0.73	0.80
letter			1600	32	6.25	0.61	0.92	0.59	0.55	0.75	0.77
musk			3062	166	3.17	1.00	0.60	1.00	0.96	1.00	1.00
annthyroid			7200	6	7.42	0.80	0.71	0.61	0.78	0.68	0.84
satimage			5803	36	1.22	0.99	0.34	0.97	0.96	0.98	0.99
thyroid			3772	6	2.47	0.98	0.48	0.95	0.98	0.98	0.98
vowels			1456	12	3.43	0.69	0.93	0.66	0.59	0.59	0.91
waveform			3443	21	2.90	0.73	0.74	0.69	0.59	0.83	0.76
wbc			223	9	4.48	1.00	0.92	0.99	1.00	0.99	1.00
wdbc			367	30	2.72	0.99	0.98	0.99	0.97	1.00	0.99
wilt			4819	5	5.33	0.46	0.69	0.41	0.39	0.34	0.53
aid	categorical		4278	114	1.40	0.65	0.58	0.66	0.66	0.61	0.64
apascal			12694	64	1.39	0.49	0.55	0.66	0.66	0.55	0.56
cmc			1472	8	1.97	0.57	0.51	0.59	0.59	0.53	0.57
reuters			12896	100	1.84	0.98	0.95	0.99	0.99	0.98	0.98
solarflare			1065	11	4.04	0.80	0.55	0.84	0.84	0.85	0.81
ncii	graph		2160	1	4.77	0.48	0.56	0.46	0.49	0.47	0.53
aids			1680	1	4.76	0.92	0.83	0.96	0.92	0.99	0.99
enzymes			105	1	4.76	0.76	0.61	0.68	0.72	0.63	0.63
proteins			696	1	4.47	0.54	0.58	0.35	0.67	0.68	0.70
dd			726	1	4.82	0.66	0.46	0.31	0.75	0.79	0.78
earthquakes	complex	time series	387	512	4.91	0.61	0.57	0.49	0.56	0.43	0.64
aibo			367	70	4.90	0.50	0.63	0.50	0.46	0.55	0.55
ECGFiveDays			465	136	4.95	0.80	0.91	0.75	0.67	0.74	0.79
MPOC			583	80	4.97	0.68	0.75	0.62	0.53	0.66	0.61
amazon	text		10000	768	5.00	0.52	0.55	0.51	0.52	0.49	0.50
imdb			10000	768	5.00	0.47	0.52	0.47	0.47	0.48	0.50
yelp			10000	768	5.00	0.54	0.59	0.55	0.56	0.50	0.54
cifar	image		5263	512	5.00	0.73	0.73	0.68	0.71	0.68	0.71
fashionmnist			6315	512	5.00	0.84	0.74	0.76	0.83	0.81	0.83
svhn			5208	512	5.00	0.56	0.66	0.48	0.54	0.57	0.55
items	mixed	sequences of sets	210	1	4.76	0.83	0.83	0.84	0.84	0.75	0.76
length			210	1	4.76	0.85	0.87	0.92	0.92	0.87	0.81
order			210	1	4.76	0.53	0.53	0.55	0.59	0.51	0.54
ovarian	multiomics		125	50	4.80	0.50	0.29	0.45	0.57	0.33	0.69
breast			770	50	3.64	0.62	0.83	0.49	0.63	0.83	0.83
rosmap			177	600	4.52	0.62	0.60	0.68	0.67	0.70	0.66
Avg. rank						3.50	3.57	3.78	3.51	3.78	2.85