

Supplementary materials

Andrea Di Via ^{†1}, Roberto Di Via² and Ulderico Fugacci³

What follows includes comprehensive tables from the paper, featuring additional samples experiments. The best results for each metric are highlighted in bold. In addition, the deep learning plots for each dataset have been represented as bar plots instead of line charts to improve the readability.

1. Deep learning plots

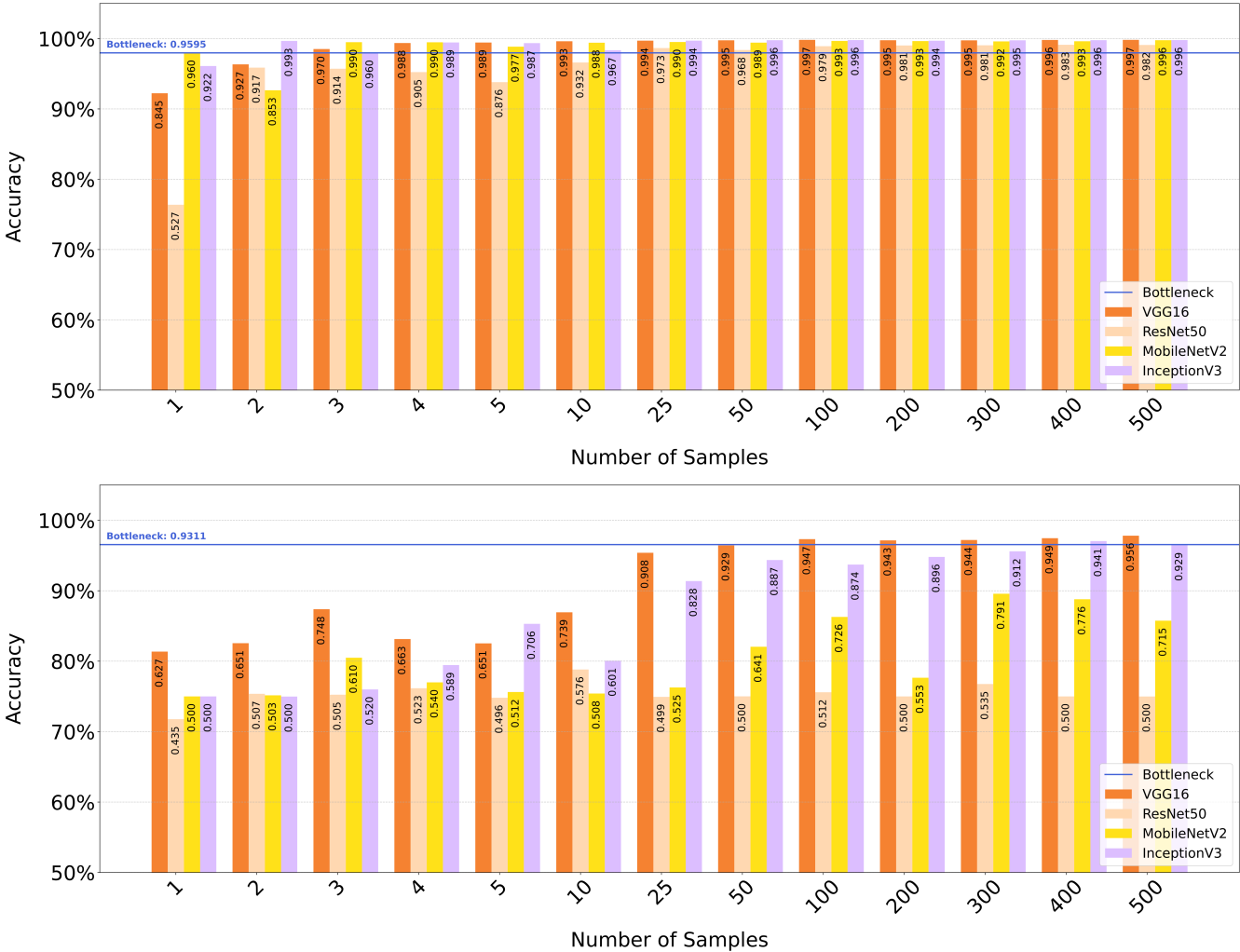


Figure 1: Bar chart illustrating the accuracy comparison between the Bottleneck method and deep learning models across varying numbers of training samples. Above is the Cracks dataset, and below is the Malaria dataset.

2. Deep learning performances for both Cracks and Malaria dataset

Samples	Method	Cracks dataset				Malaria dataset			
		Acc.	Prec.	Rec.	F1	Acc.	Prec.	Rec.	F1
1	VGG16	0.8446	0.9975	0.6910	0.8164	0.6272	0.7203	0.4158	0.5273
	ResNet50	0.5274	0.9294	0.0592	0.1114	0.4354	0.3976	0.2507	0.3075
	MobileNetV2	0.9596	0.9952	0.9238	0.9581	0.5000	0.0000	0.0000	0.0000
	InceptionV3	0.9221	0.9994	0.8448	0.9156	0.5000	0.5000	0.0004	0.0007
2	VGG16	0.9268	0.9991	0.8542	0.9210	0.6511	0.7198	0.4949	0.5865
	ResNet50	0.9175	0.9997	0.8352	0.9101	0.5074	0.5040	0.9380	0.6557
	MobileNetV2	0.8531	0.9982	0.7075	0.8281	0.5031	0.5044	0.3520	0.4146
	InceptionV3	0.9934	0.9972	0.9895	0.9933	0.4996	0.4706	0.0058	0.0115
3	VGG16	0.9704	0.9813	0.9590	0.9700	0.7476	0.8794	0.5740	0.6946
	ResNet50	0.9144	1.0000	0.8288	0.9064	0.5049	0.5616	0.0446	0.0827
	MobileNetV2	0.9899	0.9905	0.9892	0.9899	0.6098	0.6120	0.5998	0.6058
	InceptionV3	0.9595	0.9984	0.9205	0.9579	0.5198	0.5126	0.8030	0.6258
4	VGG16	0.9875	0.9947	0.9802	0.9874	0.6629	0.6061	0.9303	0.7340
	ResNet50	0.9048	1.0000	0.8095	0.8947	0.5232	0.6151	0.1241	0.2065
	MobileNetV2	0.9896	0.9868	0.9925	0.9897	0.5399	0.5915	0.2580	0.3593
	InceptionV3	0.9891	0.9895	0.9888	0.9891	0.5889	0.6259	0.4419	0.5181
5	VGG16	0.9889	0.9952	0.9825	0.9888	0.6506	0.6123	0.8211	0.7015
	ResNet50	0.8762	1.0000	0.7525	0.8588	0.4962	0.4981	0.9844	0.6615
	MobileNetV2	0.9771	0.9704	0.9842	0.9773	0.5125	0.7226	0.0406	0.0769
	InceptionV3	0.9872	0.9974	0.9770	0.9871	0.7061	0.7363	0.6422	0.6860
10	VGG16	0.9925	0.9982	0.9868	0.9925	0.7388	0.8142	0.6187	0.7031
	ResNet50	0.9321	0.9997	0.8645	0.9272	0.5764	0.5619	0.6930	0.6206
	MobileNetV2	0.9884	0.9853	0.9915	0.9884	0.5082	0.7922	0.0221	0.0431
	InceptionV3	0.9671	0.9387	0.9995	0.9682	0.6014	0.6206	0.5218	0.5669
25	VGG16	0.9942	0.9992	0.9892	0.9942	0.9078	0.9599	0.8512	0.9023
	ResNet50	0.9732	0.9971	0.9492	0.9726	0.4987	0.4720	0.0214	0.0410
	MobileNetV2	0.9904	0.9871	0.9938	0.9904	0.5254	0.7134	0.0849	0.1518
	InceptionV3	0.9942	0.9952	0.9932	0.9942	0.8275	0.8302	0.8233	0.8267

Table 1: Performance metrics for deep learning models for the Cracks and Malaria datasets across various number of samples per class ranging from 1 to 25.

Samples	Method	Cracks dataset				Malaria dataset			
		Acc.	Prec.	Rec.	F1	Acc.	Prec.	Rec.	F1
50	VGG16	0.9952	0.9965	0.9940	0.9952	0.9287	0.9084	0.9536	0.9304
	ResNet50	0.9676	0.9973	0.9378	0.9666	0.5002	0.5001	0.9989	0.6665
	MobileNetV2	0.9886	0.9897	0.9875	0.9886	0.6410	0.6462	0.6230	0.6344
	InceptionV3	0.9955	0.9945	0.9965	0.9955	0.8872	0.8949	0.8774	0.8860
100	VGG16	0.9965	0.9985	0.9945	0.9965	0.9465	0.9624	0.9292	0.9455
	ResNet50	0.9785	0.9948	0.9620	0.9781	0.5118	0.5337	0.1869	0.2768
	MobileNetV2	0.9934	0.9920	0.9948	0.9934	0.7257	0.9430	0.4804	0.6365
	InceptionV3	0.9959	0.9955	0.9962	0.9959	0.8743	0.8799	0.8668	0.8733
200	VGG16	0.9954	0.9945	0.9962	0.9954	0.9430	0.9685	0.9158	0.9414
	ResNet50	0.9806	0.9921	0.9690	0.9804	0.5000	0.5000	1.0000	0.6667
	MobileNetV2	0.9930	0.9893	0.9968	0.9930	0.5530	0.9966	0.1063	0.1921
	InceptionV3	0.9940	0.9933	0.9948	0.9940	0.8960	0.8905	0.9031	0.8968
300	VGG16	0.9951	0.9945	0.9958	0.9951	0.9441	0.9356	0.9539	0.9447
	ResNet50	0.9811	0.9931	0.9690	0.9809	0.5352	0.6054	0.2021	0.3030
	MobileNetV2	0.9921	0.9886	0.9958	0.9922	0.7915	0.9763	0.5976	0.7414
	InceptionV3	0.9952	0.9918	0.9988	0.9953	0.9120	0.9192	0.9035	0.9113
400	VGG16	0.9962	0.9940	0.9985	0.9963	0.9492	0.9602	0.9372	0.9486
	ResNet50	0.9825	0.9906	0.9742	0.9824	0.5000	0.0000	0.0000	0.0000
	MobileNetV2	0.9925	0.9920	0.9930	0.9925	0.7759	0.9929	0.5559	0.7127
	InceptionV3	0.9956	0.9945	0.9968	0.9956	0.9410	0.9606	0.9198	0.9398
500	VGG16	0.9965	0.9955	0.9975	0.9965	0.9563	0.9708	0.9409	0.9556
	ResNet50	0.9820	0.9876	0.9762	0.9819	0.4998	0.4999	0.9996	0.6665
	MobileNetV2	0.9955	0.9935	0.9975	0.9955	0.7150	0.9837	0.4372	0.6054
	InceptionV3	0.9959	0.9958	0.9960	0.9959	0.9291	0.9368	0.9202	0.9284

Table 2: Performance metrics for deep learning models for the Cracks and Malaria datasets across various number of samples per class ranging from 50 to 500.

3. Machine learning (w/wo Persistent homology) for the Cracks dataset

Samples	Method	Cracks dataset							
		ML				persIm ML			
		Acc.	Prec.	Rec.	F1	Acc.	Prec.	Rec.	F1
1	Log Reg	0.6940	0.6999	0.6792	0.6894	0.9140	0.9952	0.8320	0.9063
	SVM	0.6940	0.6999	0.6792	0.6894	0.9140	0.9952	0.8320	0.9063
	RF	0.5919	0.6284	0.4498	0.5243	0.7220	1.0000	0.4440	0.6150
5	Log Reg	0.7522	0.8171	0.6500	0.7240	0.9440	0.9625	0.9240	0.9429
	SVM	0.7704	0.7299	0.8585	0.7890	0.9600	0.9528	0.9680	0.9603
	RF	0.6738	0.7395	0.5365	0.6218	0.8800	0.9750	0.7800	0.8667
10	Log Reg	0.7492	0.7726	0.7065	0.7381	0.9280	0.9908	0.8640	0.9231
	SVM	0.8080	0.8596	0.7362	0.7932	0.9780	0.9723	0.9840	0.9781
	RF	0.7151	0.7318	0.6792	0.7045	0.4960	0.4980	0.9880	0.6622
25	Log Reg	0.7562	0.7889	0.6998	0.7417	0.9540	0.9595	0.9480	0.9537
	SVM	0.8635	0.8444	0.8912	0.8672	0.9620	0.9358	0.9920	0.9631
	RF	0.7061	0.6917	0.7438	0.7168	0.9660	0.9569	0.9760	0.9663
50	Log Reg	0.8028	0.8421	0.7452	0.7907	0.9680	0.9570	0.9800	0.9684
	SVM	0.8966	0.8959	0.8975	0.8967	0.972	0.9574	0.988	0.9724
	RF	0.7569	0.7510	0.7685	0.7597	0.9680	0.9570	0.9800	0.9684
100	Log Reg	0.8336	0.8919	0.7592	0.8203	0.9680	0.9606	0.9760	0.9683
	SVM	0.9430	0.9847	0.9000	0.9404	0.9700	0.9572	0.9840	0.9704
	RF	0.8934	0.9319	0.8488	0.8884	0.9680	0.9570	0.9800	0.9684
200	Log Reg	0.8669	0.9467	0.7775	0.8538	0.9740	0.9798	0.9680	0.9738
	SVM	0.9559	0.9810	0.9298	0.9547	0.9760	0.9798	0.9720	0.9759
	RF	0.9461	0.9832	0.9078	0.9440	0.9760	0.9648	0.9880	0.9763
300	Log Reg	0.8459	0.8918	0.7872	0.8363	0.9576	0.9607	0.9542	0.9575
	SVM	0.9549	0.9822	0.9265	0.9536	0.9654	0.9600	0.9712	0.9656
	RF	0.9296	0.9280	0.9315	0.9298	0.9648	0.9523	0.9785	0.9652
400	Log Reg	0.8721	0.9288	0.8060	0.8631	0.9616	0.9622	0.9610	0.9616
	SVM	0.9586	0.9784	0.9380	0.9578	0.9674	0.9574	0.9782	0.9677
	RF	0.9596	0.9698	0.9488	0.9592	0.9620	0.9538	0.9710	0.9623
500	Log Reg	0.8734	0.9290	0.8085	0.8646	0.9614	0.9547	0.9688	0.9617
	SVM	0.9582	0.9801	0.9355	0.9573	0.9670	0.9532	0.9822	0.9675
	RF	0.9668	0.9799	0.9530	0.9663	0.9585	0.9526	0.9650	0.9588

Table 3: Performance metrics for machine learning models for the Cracks dataset across various number of samples per class, applied to raw and persistence images.

4. Machine learning performances for the Malaria dataset

Samples	Method	Malaria dataset							
		ML				persIm ML			
		Acc.	Prec.	Rec.	F1	Acc.	Prec.	Rec.	F1
1	Log Reg	0.3881	0.4007	0.4517	0.4247	0.9087	0.9581	0.8549	0.9035
	SVM	0.3881	0.4007	0.4517	0.4247	0.9089	0.9574	0.8560	0.9038
	RF	0.3975	0.4087	0.4590	0.4324	0.6063	0.9508	0.2242	0.3629
5	Log Reg	0.6020	0.6200	0.5269	0.5696	0.8959	0.9576	0.8284	0.8883
	SVM	0.6009	0.6161	0.5352	0.5728	0.9272	0.9308	0.9231	0.9269
	RF	0.6216	0.6144	0.6528	0.6330	0.8641	0.9593	0.7605	0.8484
10	Log Reg	0.5709	0.5821	0.5029	0.5396	0.9082	0.9518	0.8599	0.9035
	SVM	0.5776	0.6192	0.4035	0.4886	0.9311	0.9314	0.9307	0.931
	RF	0.5923	0.5900	0.6052	0.5975	0.9151	0.9525	0.8737	0.9114
25	Log Reg	0.5847	0.5888	0.5617	0.5749	0.8988	0.9541	0.8378	0.8922
	SVM	0.6272	0.6699	0.5015	0.5736	0.9311	0.9314	0.9307	0.931
	RF	0.6477	0.6709	0.5798	0.6220	0.9156	0.9540	0.8734	0.9119
50	Log Reg	0.6054	0.6097	0.5860	0.5976	0.9231	0.9397	0.9042	0.9216
	SVM	0.5971	0.6322	0.4641	0.5353	0.9311	0.9314	0.9307	0.9310
	RF	0.6139	0.6098	0.6328	0.6211	0.9111	0.9572	0.8607	0.9064
100	Log Reg	0.6306	0.6343	0.6168	0.6255	0.9066	0.9491	0.8592	0.9019
	SVM	0.6549	0.6545	0.6564	0.6554	0.9311	0.9314	0.9307	0.931
	RF	0.6798	0.6981	0.6335	0.6643	0.9165	0.9505	0.8788	0.9133
200	Log Reg	0.6179	0.6270	0.5824	0.6038	0.9142	0.9499	0.8745	0.9106
	SVM	0.6646	0.6679	0.6546	0.6612	0.9307	0.9332	0.9278	0.9305
	RF	0.6856	0.7007	0.6480	0.6733	0.9260	0.9437	0.9060	0.9245
300	Log Reg	0.6226	0.6327	0.5849	0.6078	0.9242	0.9452	0.9006	0.9223
	SVM	0.6704	0.6667	0.6814	0.6740	0.9309	0.9317	0.9300	0.9308
	RF	0.6836	0.6869	0.6749	0.6808	0.9251	0.9483	0.8991	0.9231
400	Log Reg	0.6313	0.6360	0.6143	0.6250	0.9189	0.9415	0.8933	0.9168
	SVM	0.6782	0.6783	0.6777	0.6780	0.9312	0.9325	0.9282	0.9304
	RF	0.6877	0.6935	0.6628	0.6778	0.9244	0.9479	0.8985	0.9222
500	Log Reg	0.6136	0.6220	0.5791	0.5998	0.9224	0.9429	0.8991	0.9205
	SVM	0.6711	0.6597	0.7068	0.6824	0.9300	0.9322	0.9274	0.9298
	RF	0.7037	0.7032	0.7050	0.7041	0.9276	0.9416	0.9118	0.9265

Table 4: Performance metrics for machine learning models for the Malaria dataset across various number of samples per class, applied to raw and persistence images.