Machine Learning - Assignment 3

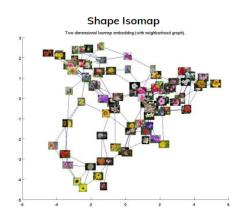
Maor Sagi

Introduction

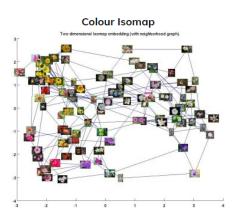
In this assignment we requested to practice the usage of Convolutional Neural Network (CNN) and Transfer Learning. The task is flower classification by using 2 different pre-trained models.

Data and Preprocessing

The data used for this assignment is the flowers dataset (LINK), containing 102 different types and a total number of 8189 images. The maximum number of images in a class is 258, the minimum number is 40 and The average number of images in each class is 80. The flowers are characterized by different colors and different shapes.



and then I sampled from each class randomly.



An Isomap of categories in the dataset, randomly sampled from the category. Image taken from here.

All the images resized to 256 x 256 pixels. To improve generalization of the model, I applied several augmentation techniques - rescaling, zooming, rotations, width and heights shifts, horizontal and vertical flipping and shearing (distorted along an axis). The augmentations adjustments generated randomly per image, using Image Data Generator. In this assignment we requested to split the data to 50% training set, 25% validation and 25% test set. I splitted the data balance split, meaning, first I categorized all images by class

Architecture and Pretrained Models

After Trying different pre-train models with different configurations, the models performed best are - MobileNet and InceptionV3 using ImageNet weights. I wrapped these models with extra hidden layers, for each model I tested different variations till proceeding to the best results.

1st model - based on Inception V3 pretrained model:

- 1. Inception V3 layers
- 2. Global Average Pooling 2D
- 3. Batch Normalization
- 4. Dropout (0.3 rate)
- 5. Dense 256 and ReLU
- 6. Batch Normalization
- 7. Dense 102 and Softmax

2nd model - based on Mobile Net pretrained model:

- 1. Mobile Net layers
- 2. Global Average Pooling 2D
- 3. Dense 2048 and ReLU
- 4. Batch Normalization
- 5. Dense 1024 and ReLU
- 6. Batch Normalization
- 7. Dense 1024 and ReLU
- 8. Batch Normalization
- 9. Dropout (0.2 rate)
- 10. Dense 512 and ReLU
- 11. Dropout (0.2 rate)
- 12. Batch Normalization
- 13. Dropout (0.1 rate)
- 14. Dense 256 and ReLU
- 15. Batch Normalization
- 16. Dense 102 and Softmax

Both models compiled with Adam optimizer, learning rate of 0.001. The loss function was defined as categorical cross entropy.

First The models trained while all the pretrained models' weights were frozen and later I commited fine tuning by unfreezing the weights of all layers, except Batch Normalization who commonly recommended to be kept on inference mode to avoid unexpected performance on the validation or test set, as described in <u>Fahdi Kanavati et al. (2021)</u>. Later I train the model for another session. In the second train, the models compiled with a smaller learning rate, 1e-4.

Results and Discussion

The 1st and 2nd models experiments configurations:

Initial training:

- 1. Frozen weights for the pretrained model layers.
- 2. 30 epochs.
- 3. Best model checkpoint callback, monitored by minimum validation loss.
- 4. Split of train-test was according to the given split, otherwise by stratified split while test size is 20%.
- 5. Reduce learning rate callback when validation loss stops improving with patience factor of 2 epochs.

Secondary training:

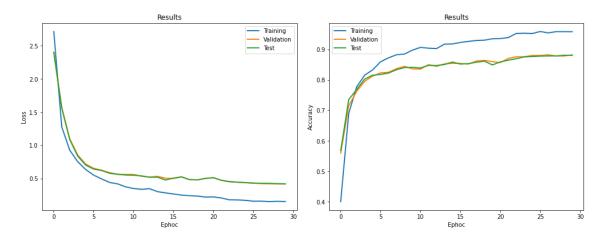
- 1. Unfreeze frozen weights except Batch Normalization layers.
- 2. 60 epochs.
- 3. Best model checkpoint callback, monitored by minimum validation loss.
- 4. Early stopping callback, monitored by minimum validation loss with patience factor of 4 epochs.
- 5. Reduce learning rate callback when validation loss stops improving with patience factor of 2 epochs.

I will discuss the results for each experiment, I splitted the dataset randomly as described above. I executed the experiment with the above configuration twice for each architecture, 4 executions in total, 4 different data separations. I will discuss the results as presented in the figures, the detailed results can be found on the appendices, and attached to the submission file together with the exact data splits.

The initial epoch in the secondary training is defined to be the last result on the initial training, that is explaining the starting point in the secondary loss and accuracy figures.

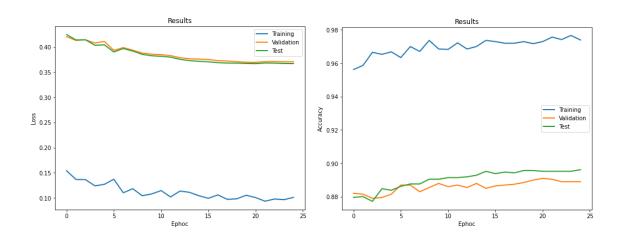
1st experiment - based on Inception V3 pretrained model:

1st execution:



loss: 0.4216 accuracy: 0.8801

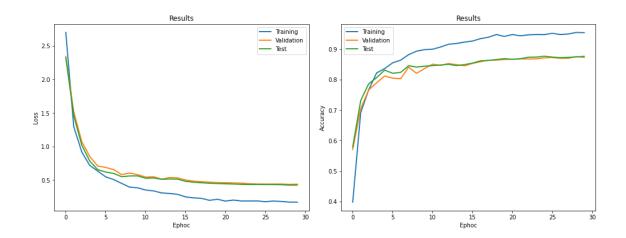
2nd execution:



loss: 0.3672 accuracy: 0.8962

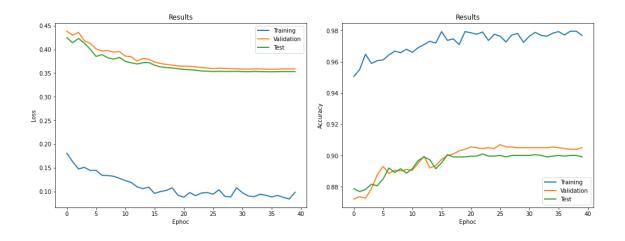
2nd experiment - based on Inception V3 pretrained model:

1st execution:



loss: 0.4258 accuracy: 0.8763

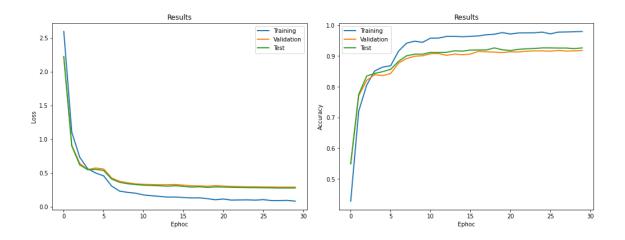
2nd execution:



loss: 0.3531 accuracy: 0.8991

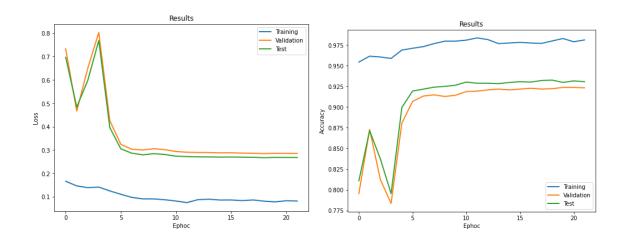
3rd experiment - based on Mobile Net pretrained model:

1st execution:



loss: 0.2782 accuracy: 0.9261

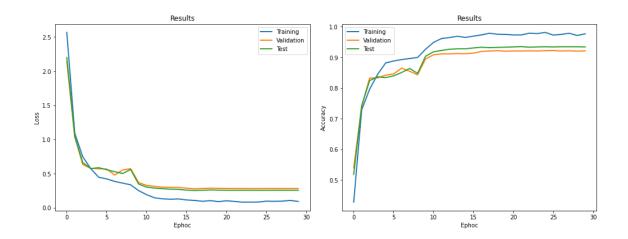
2nd execution:



loss: 0.2683 accuracy: 0.9308

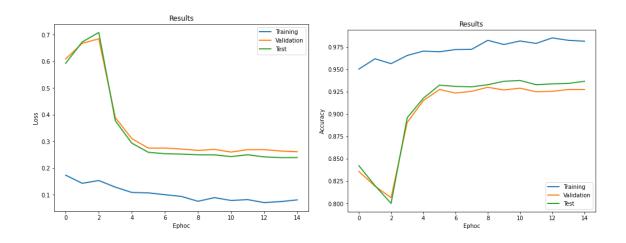
4th experiment - based on Mobile Net pretrained model:

1st execution:



loss: 0.2550 accuracy: 0.9341

2nd execution:



loss: 0.2389 accuracy: 0.9365

We can see for all graphs the validation set and the test set provide similar results. The Mobile Net based model provides best performance of 93.65% accuracy. The Inception V3 accuracy is pretty high as well, 89.9%.

Appendices

Detailed Results:

1st experiment - based on Inception V3 pretrained model:

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.400737	2.712055	0.559482	2.399103	0.567773	2.397126
1	0.689435	1.276012	0.714783	1.571921	0.735071	1.557508
2	0.776904	0.926258	0.762071	1.09801	0.768246	1.085515
3	0.815233	0.754019	0.794923	0.85469	0.80237	0.837704
4	0.832678	0.635891	0.812344	0.717681	0.815166	0.702521
5	0.859214	0.551524	0.822797	0.653773	0.817536	0.642973
6	0.872236	0.494393	0.824788	0.625341	0.822275	0.621462
7	0.88231	0.441687	0.836735	0.589079	0.833175	0.579493
8	0.884521	0.420462	0.844201	0.563071	0.840284	0.564843
9	0.897543	0.375722	0.835739	0.562691	0.840758	0.553542
10	0.906388	0.350401	0.835241	0.562583	0.838863	0.550453
11	0.903686	0.338832	0.849676	0.536213	0.847393	0.540068
12	0.902457	0.348001	0.844201	0.522404	0.846919	0.520517
13	0.916953	0.303356	0.852165	0.533768	0.850237	0.521175
14	0.91769	0.285041	0.854156	0.50505	0.858294	0.478574

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15	0.922604	0.26818	0.853659	0.501664	0.851659	0.506993
16	0.926044	0.252452	0.851668	0.526492	0.853081	0.526229
17	0.928993	0.241928	0.861623	0.485115	0.85782	0.484006
18	0.929975	0.236808	0.863614	0.481055	0.861137	0.480294
19	0.934644	0.221979	0.859632	0.503469	0.848815	0.502795
20	0.935381	0.224885	0.856645	0.51304	0.858768	0.512957
21	0.938575	0.208199	0.870085	0.475029	0.864455	0.473584
22	0.952088	0.180675	0.87556	0.454291	0.869194	0.454085
23	0.952826	0.179105	0.87556	0.44556	0.875355	0.446407
24	0.951843	0.173241	0.88004	0.438484	0.876303	0.440486
25	0.958477	0.159819	0.88004	0.430414	0.877725	0.432749
26	0.953808	0.16055	0.882031	0.424478	0.878199	0.428995
27	0.957985	0.154252	0.878049	0.421957	0.878199	0.428403
28	0.957985	0.158308	0.878049	0.419439	0.880569	0.424522
29	0.95774	0.155464	0.882031	0.417957	0.880095	0.421577

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.956265	0.15394	0.882031	0.420655	0.879621	0.424842
1	0.958722	0.136614	0.881533	0.413162	0.880095	0.414122
2	0.966585	0.136309	0.879044	0.414334	0.877251	0.414468
3	0.965356	0.123929	0.879542	0.407992	0.884834	0.403501

4	0.96683	0.126935	0.881533	0.411045	0.883886	0.404393
5	0.963391	0.137124	0.887008	0.393333	0.886256	0.390175
6	0.970025	0.110178	0.887008	0.39889	0.887678	0.397109
7	0.967076	0.118288	0.883026	0.393698	0.887678	0.392049
8	0.97371	0.104289	0.885515	0.388063	0.890521	0.385415
9	0.96855	0.107782	0.888004	0.385784	0.890521	0.382664
10	0.968305	0.114551	0.886013	0.384824	0.891469	0.38157
11	0.972236	0.101831	0.887008	0.383014	0.891469	0.379855
12	0.96855	0.113642	0.885515	0.37877	0.891943	0.37555
13	0.970025	0.111024	0.888004	0.376595	0.892891	0.372999
14	0.97371	0.104242	0.885017	0.375935	0.895261	0.371557
15	0.972973	0.099184	0.886511	0.375186	0.893839	0.370643
16	0.97199	0.10594	0.887008	0.372975	0.894787	0.368932
17	0.97199	0.097085	0.887506	0.372205	0.894313	0.36815
18	0.972973	0.098124	0.888502	0.370892	0.895735	0.368038
19	0.971744	0.105324	0.889995	0.3698	0.895735	0.367447
20	0.972973	0.100557	0.890991	0.369551	0.895261	0.36701
21	0.975676	0.093401	0.890493	0.371117	0.895261	0.368245
22	0.974202	0.097818	0.889	0.371239	0.895261	0.368019
23	0.976658	0.0964	0.889	0.370666	0.895261	0.367392
24	0.973956	0.100991	0.889	0.370798	0.896209	0.367153

2nd experiment - based on Inception V3 pretrained model:

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.398034	2.697731	0.570931	2.340134	0.579147	2.32041
1	0.6914	1.294044	0.704828	1.513701	0.73128	1.460068
2	0.766585	0.91877	0.766053	1.07297	0.786256	1.020844
3	0.822113	0.718556	0.789945	0.850488	0.806635	0.783481
4	0.836118	0.633733	0.811847	0.70824	0.83128	0.654625
5	0.855037	0.547039	0.804878	0.687777	0.820853	0.618884
6	0.863882	0.507135	0.802887	0.654819	0.823697	0.598013
7	0.881818	0.449723	0.840717	0.581431	0.845498	0.549997
8	0.893366	0.393901	0.820806	0.603586	0.841232	0.560856
9	0.89828	0.384419	0.836237	0.580729	0.844076	0.561233
10	0.899754	0.351368	0.850672	0.543192	0.845972	0.526958
11	0.90688	0.338053	0.84669	0.548566	0.847867	0.531183
12	0.915971	0.308009	0.852663	0.513298	0.850237	0.511224
13	0.918673	0.300018	0.849179	0.536685	0.845972	0.51511
14	0.923096	0.287055	0.845197	0.532392	0.849289	0.512903
15	0.926536	0.249431	0.853161	0.500541	0.854028	0.481077
16	0.934398	0.234458	0.858636	0.48351	0.861611	0.467439
17	0.939066	0.227147	0.863614	0.475228	0.863033	0.459246

18	0.947912	0.197722	0.864111	0.468399	0.865877	0.45201
19	0.941769	0.213602	0.866103	0.462511	0.86872	0.44769
20	0.947912	0.186321	0.867098	0.460409	0.866825	0.443852
21	0.943735	0.201741	0.868591	0.4574	0.868246	0.4406
22	0.946929	0.187815	0.867596	0.453558	0.87346	0.435994
23	0.947912	0.187961	0.868094	0.444896	0.873934	0.433165
24	0.947666	0.18867	0.871578	0.440665	0.876777	0.43422
25	0.951843	0.17773	0.872573	0.440027	0.873934	0.43238
26	0.947912	0.187371	0.870085	0.439919	0.872038	0.432873
27	0.949631	0.181064	0.870085	0.440817	0.872986	0.43066
28	0.954791	0.171585	0.87556	0.435462	0.874408	0.424191
29	0.954054	0.170666	0.873071	0.438438	0.876303	0.425752

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.950614	0.180745	0.872076	0.438356	0.878673	0.424863
1	0.955037	0.162729	0.873569	0.429798	0.876777	0.414013
2	0.964865	0.147428	0.872573	0.436069	0.878199	0.423203
3	0.958968	0.150808	0.878547	0.418137	0.881517	0.413023
4	0.960688	0.144146	0.887506	0.412594	0.880569	0.400308
5	0.961179	0.144569	0.892982	0.400996	0.884834	0.385117
6	0.964373	0.133944	0.888502	0.396743	0.891943	0.388983

0.96683	0.133429	0.890493	0.397336	0.8891	0.382167
0.965848	0.131953	0.889995	0.394107	0.891469	0.379447
0.968059	0.127609	0.890991	0.395203	0.888626	0.382902
0.966093	0.122718	0.890493	0.385839	0.891469	0.374617
0.969042	0.118892	0.894973	0.384485	0.896682	0.371746
0.971007	0.109666	0.899452	0.375245	0.899052	0.369128
0.973219	0.105736	0.891986	0.380811	0.897156	0.371704
0.97199	0.10897	0.893479	0.379089	0.891469	0.372226
0.979361	0.095657	0.897461	0.373304	0.895261	0.36658
0.97371	0.099673	0.89995	0.370123	0.900474	0.362973
0.974693	0.102036	0.900946	0.368233	0.899052	0.361586
0.971007	0.107542	0.902937	0.366593	0.899052	0.360801
0.979361	0.091848	0.903932	0.364927	0.899052	0.359124
0.978624	0.087784	0.905426	0.364628	0.899526	0.35766
0.977641	0.097547	0.904928	0.364219	0.899526	0.357098
0.979115	0.090945	0.90443	0.363205	0.900948	0.355847
0.97371	0.096511	0.904928	0.361618	0.899526	0.354448
0.977641	0.097472	0.90443	0.360587	0.899526	0.353919
0.976413	0.094028	0.906919	0.359291	0.9	0.353168
0.972727	0.103399	0.905426	0.360301	0.899052	0.35362
0.97715	0.089547	0.905426	0.359721	0.9	0.353362
0.978133	0.088137	0.904928	0.359051	0.9	0.353358
0.972482	0.107668	0.904928	0.358862	0.9	0.353613
	0.965848 0.968059 0.966093 0.969042 0.971007 0.973219 0.97371 0.97371 0.974693 0.974693 0.974693 0.977641 0.977641 0.977641 0.977641 0.977641 0.977641 0.977641 0.9777641	0.965848	0.965848 0.131953 0.889995 0.968059 0.127609 0.890991 0.966093 0.122718 0.890493 0.969042 0.118892 0.894973 0.971007 0.109666 0.899452 0.973219 0.105736 0.891986 0.979361 0.095657 0.897461 0.97371 0.099673 0.89995 0.974693 0.102036 0.900946 0.979361 0.091848 0.903932 0.978624 0.087784 0.905426 0.977641 0.097547 0.904928 0.97371 0.096511 0.904928 0.977641 0.097472 0.90443 0.977641 0.097472 0.90443 0.977641 0.097472 0.90443 0.977641 0.097472 0.90443 0.977641 0.094028 0.906919 0.97715 0.089547 0.905426 0.978133 0.088137 0.904928	0.965848 0.131953 0.889995 0.394107 0.968059 0.127609 0.890991 0.395203 0.966093 0.122718 0.890493 0.385839 0.969042 0.118892 0.894973 0.384485 0.971007 0.109666 0.899452 0.375245 0.973219 0.105736 0.891986 0.380811 0.97199 0.10897 0.893479 0.379089 0.97371 0.099673 0.89995 0.370123 0.974693 0.102036 0.900946 0.368233 0.971007 0.107542 0.902937 0.366593 0.979361 0.091848 0.903932 0.364927 0.978624 0.087784 0.905426 0.36428 0.977641 0.097547 0.904928 0.361618 0.9777641 0.097472 0.90443 0.360587 0.9776413 0.094028 0.906919 0.359291 0.97715 0.089547 0.905426 0.360301 0.97715 0.089547 0.905426<	0.965848 0.131953 0.889995 0.394107 0.891469 0.968059 0.127609 0.890991 0.395203 0.888626 0.966093 0.122718 0.890493 0.385839 0.891469 0.969042 0.118892 0.894973 0.384485 0.896682 0.971007 0.109666 0.899452 0.375245 0.899052 0.973219 0.105736 0.891986 0.380811 0.897156 0.97199 0.10897 0.893479 0.379089 0.891469 0.979361 0.095657 0.897461 0.373304 0.895261 0.97371 0.099673 0.89995 0.370123 0.900474 0.974693 0.102036 0.900946 0.368233 0.899052 0.979361 0.091848 0.903932 0.364927 0.8999526 0.9778624 0.087784 0.905426 0.364219 0.899526 0.977641 0.097547 0.904928 0.361618 0.899526 0.9777641 0.097472 0.90443 0.360587 </td

30	0.976413	0.09721	0.904928	0.358191	0.9	0.353089
31	0.97887	0.090313	0.904928	0.358169	0.900474	0.352732
32	0.976904	0.088705	0.904928	0.358664	0.9	0.353499
33	0.976413	0.093961	0.904928	0.358583	0.899052	0.35313
34	0.978378	0.091782	0.905426	0.358147	0.899526	0.352738
35	0.979361	0.088228	0.904928	0.358121	0.9	0.352641
36	0.97715	0.091532	0.90443	0.358134	0.899526	0.352809
37	0.979607	0.087371	0.903932	0.358716	0.9	0.353281
38	0.979607	0.083903	0.903932	0.35847	0.9	0.353016
39	0.976904	0.09816	0.904928	0.358605	0.899052	0.353091

3rd experiment - based on Mobile Net pretrained model:

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.427764	2.594738	0.549527	2.211029	0.549289	2.223706
1	0.720147	1.104411	0.770035	0.911722	0.774882	0.900108
2	0.80516	0.732204	0.820806	0.642602	0.834597	0.619907
3	0.851106	0.561439	0.838726	0.549873	0.843128	0.548008
4	0.863636	0.499356	0.836237	0.573468	0.848815	0.553569
5	0.868059	0.456784	0.842708	0.558858	0.856398	0.53687
6	0.915971	0.306097	0.876556	0.426757	0.882464	0.40922

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41523	0.231155	0.891488	0.376043	0.900474	0.362085
47912	0.21297	0.898955	0.354533	0.905687	0.340478
44226	0.200864	0.900448	0.338591	0.905687	0.328887
5774	0.174518	0.906919	0.331984	0.911374	0.319302
57985	0.163201	0.907914	0.329612	0.9109	0.31576
63391	0.153504	0.901941	0.326451	0.911848	0.310473
63636	0.142722	0.905923	0.326242	0.916588	0.30509
62408	0.14394	0.903932	0.330031	0.91564	0.311547
63636	0.137386	0.905923	0.319529	0.918957	0.301883
65111	0.130488	0.914883	0.310144	0.919431	0.290691
68796	0.132232	0.91339	0.309271	0.919431	0.295132
70516	0.119454	0.912394	0.303623	0.926066	0.286247
75921	0.103723	0.910901	0.313061	0.919905	0.293547
71253	0.114423	0.91339	0.305364	0.917536	0.291587
74693	0.098606	0.912892	0.301413	0.921327	0.287478
74939	0.100967	0.915381	0.29853	0.923223	0.285722
75184	0.102131	0.916376	0.295765	0.924171	0.28357
77396	0.097898	0.916376	0.294649	0.926066	0.282866
71744	0.105573	0.915381	0.293086	0.926066	0.281197
77396	0.092013	0.91787	0.292731	0.925592	0.279634
77887	0.09191	0.915879	0.290309	0.925592	0.277458
78624	0.094112	0.916376	0.290493	0.923697	0.277892
79607	0.083019	0.91787	0.290392	0.926066	0.278156
	47912 44226 5774 57985 63391 63636 62408 63636 65111 68796 70516 75921 71253 74693 74939 75184 77396 71744 77396 77887 78624	47912 0.21297 44226 0.200864 5774 0.174518 57985 0.163201 63391 0.153504 63636 0.142722 62408 0.14394 63636 0.137386 65111 0.130488 68796 0.132232 70516 0.119454 75921 0.103723 71253 0.114423 74693 0.098606 74939 0.100967 75184 0.102131 77396 0.097898 71744 0.105573 77396 0.092013 77887 0.094112	479120.212970.898955442260.2008640.90044857740.1745180.906919579850.1632010.907914633910.1535040.901941636360.1427220.905923624080.143940.903932636360.1373860.905923651110.1304880.914883687960.1322320.91339705160.1194540.912394759210.1037230.910901712530.1144230.91339746930.0986060.912892749390.1009670.915381751840.1021310.916376773960.0978980.916376717440.1055730.915381773960.0920130.91787778870.091910.915879786240.0941120.916376	47912 0.21297 0.898955 0.354533 44226 0.200864 0.900448 0.338591 5774 0.174518 0.906919 0.331984 57985 0.163201 0.907914 0.329612 63391 0.153504 0.901941 0.326451 63636 0.142722 0.905923 0.326242 62408 0.14394 0.903932 0.330031 63636 0.137386 0.905923 0.319529 65111 0.130488 0.914883 0.310144 68796 0.132232 0.91339 0.309271 70516 0.119454 0.912394 0.303623 75921 0.103723 0.910901 0.313061 71253 0.114423 0.91339 0.305364 74693 0.098606 0.912892 0.301413 74939 0.100967 0.915381 0.29853 75184 0.102131 0.916376 0.294649 71744 0.105573 0.915381 0.293086 77396 0.092013 0.91787 0.292731 77887<	47912 0.21297 0.898955 0.354533 0.905687 44226 0.200864 0.900448 0.338591 0.905687 5774 0.174518 0.906919 0.331984 0.911374 57985 0.163201 0.907914 0.329612 0.9109 63391 0.153504 0.901941 0.326451 0.911848 63636 0.142722 0.905923 0.326242 0.916588 62408 0.14394 0.903932 0.330031 0.91564 63636 0.137386 0.905923 0.319529 0.918957 65111 0.130488 0.914883 0.310144 0.919431 70516 0.119454 0.912394 0.303623 0.926066 75921 0.103723 0.910901 0.313061 0.919905 71253 0.114423 0.91339 0.305364 0.917536 74693 0.098606 0.912892 0.301413 0.921327 75184 0.102131 0.916376 0.295765 0.924171 <tr< td=""></tr<>

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.954545	0.166609	0.795421	0.732725	0.8109	0.696968
1	0.961671	0.146709	0.873071	0.466926	0.871564	0.48382
2	0.960688	0.139322	0.812344	0.649756	0.837441	0.597874
3	0.958968	0.141669	0.783474	0.80308	0.795261	0.768735
4	0.969042	0.125206	0.880538	0.426077	0.899526	0.396283
5	0.971253	0.110899	0.906919	0.324464	0.919431	0.305203
6	0.973219	0.097459	0.91339	0.303493	0.921801	0.286592
7	0.976904	0.091137	0.914883	0.300341	0.924171	0.279597
8	0.979853	0.091004	0.912892	0.305845	0.925119	0.284547
9	0.979853	0.08742	0.914385	0.301803	0.92654	0.281088
10	0.981081	0.081998	0.918865	0.293805	0.930332	0.273967
11	0.983784	0.075356	0.919363	0.290893	0.92891	0.272229
12	0.981818	0.087872	0.920856	0.289748	0.92891	0.271024
13	0.976904	0.08986	0.921852	0.289327	0.928436	0.270781
14	0.977641	0.086209	0.920856	0.288218	0.929858	0.26996
15	0.978378	0.086572	0.921852	0.288327	0.930806	0.270317
16	0.977641	0.083753	0.922847	0.287045	0.930332	0.269729
17	0.97715	0.086586	0.921852	0.286293	0.932227	0.268968

18	0.980098	0.081358	0.922349	0.284838	0.932701	0.267314
19	0.983047	0.078286	0.923843	0.286084	0.929858	0.268677
20	0.979115	0.083173	0.923843	0.285705	0.931754	0.268388
21	0.981327	0.081948	0.923345	0.285718	0.930806	0.2683

4th experiment - based on Mobile Net pretrained model:

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.42801	2.567291	0.54007	2.14001	0.518483	2.199684
1	0.729238	1.091192	0.741165	1.033818	0.742654	1.045747
2	0.79656	0.745772	0.832753	0.633602	0.824645	0.660998
3	0.844963	0.571905	0.83325	0.576488	0.836493	0.573463
4	0.881818	0.445941	0.84221	0.570823	0.834123	0.586053
5	0.888206	0.421996	0.846192	0.565698	0.83981	0.556919
6	0.892875	0.384817	0.865107	0.47815	0.850711	0.52739
7	0.896315	0.360183	0.854654	0.552471	0.863981	0.501155
8	0.899754	0.335123	0.843206	0.571839	0.847867	0.559652
9	0.926536	0.250141	0.894973	0.368924	0.903791	0.34626
10	0.948894	0.191008	0.908412	0.326267	0.918009	0.301005
11	0.961425	0.144661	0.911399	0.312437	0.922749	0.287304
12	0.964619	0.129996	0.911399	0.300833	0.92654	0.278502

13	0.968796	0.123654	0.912394	0.298248	0.927962	0.271554
14	0.965111	0.128485	0.911896	0.29639	0.927962	0.269185
15	0.969287	0.11351	0.913888	0.286661	0.930806	0.258198
16	0.973219	0.106336	0.919363	0.276382	0.933175	0.252221
17	0.978624	0.093853	0.920856	0.281348	0.931754	0.255086
18	0.97543	0.103523	0.921852	0.286381	0.932701	0.26084
19	0.974939	0.089214	0.920358	0.283408	0.933175	0.257314
20	0.973219	0.101287	0.920856	0.281725	0.934123	0.255629
21	0.973219	0.092435	0.920856	0.28056	0.935071	0.254871
22	0.97887	0.081278	0.921354	0.280984	0.933175	0.255506
23	0.977641	0.082202	0.920856	0.280139	0.934123	0.25464
24	0.981327	0.082743	0.921852	0.28002	0.934597	0.254617
25	0.972727	0.095937	0.922349	0.280911	0.934123	0.255149
26	0.974939	0.093663	0.920856	0.280933	0.934597	0.255342
27	0.978624	0.096065	0.921354	0.280483	0.934597	0.255211
28	0.971499	0.106588	0.920358	0.280906	0.934597	0.255345
29	0.976904	0.091402	0.921354	0.280541	0.934123	0.254973

epoch	accuracy	loss	val_accu racy	val_loss	test_acc uracy	test_loss
0	0.950123	0.172917	0.835739	0.609395	0.84218	0.593124
1	0.961671	0.142407	0.819313	0.667908	0.819905	0.673135

2	0.956265	0.152871	0.806371	0.684719	0.8	0.708566
3	0.965356	0.12806	0.890493	0.389155	0.895735	0.377055
4	0.97027	0.1081	0.914883	0.309831	0.917536	0.293023
5	0.969533	0.106403	0.927327	0.274624	0.932227	0.258735
6	0.97199	0.099774	0.923345	0.27476	0.930806	0.253478
7	0.972236	0.093134	0.925336	0.271096	0.930332	0.252017
8	0.98231	0.074955	0.929816	0.265922	0.932701	0.249452
9	0.977641	0.088721	0.926829	0.26971	0.936493	0.249249
10	0.981573	0.077985	0.92882	0.259752	0.937441	0.24273
11	0.97887	0.081218	0.924838	0.268699	0.932701	0.249586
12	0.985012	0.070082	0.925336	0.268831	0.933649	0.241635
13	0.98231	0.073777	0.927327	0.263614	0.934123	0.238867
14	0.981327	0.080186	0.927327	0.260941	0.936493	0.23891