

Assignment 1

COL870 : Deep Learning

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Part 1 - ResNet over Convolution Networks and different Normalization schemes

1.1 - Image Classification using Residual Network

1. Implement ResNet

The ResNet architecture as in the paper has been constructed as follows:

- A block class has been created which outputs 2 layers where each layer includes: a convolution layer, a normalization layer as specified by input from the user, and a ReLU activation layer.
- For the value of 'n' as input, the resnet class constructs a ResNet architecture with $6n+2$ layers. There are three stacks created, each of which creates 'n' number of blocks, each block has two layers - a total of $6n$ layers. This included with initial convolution layer and the final linear layer completes the desired layout.

Layer 1	Convolution + normalization + ReLU
Stack1	n blocks with 2 layers each i.e. 2n layers
Stack2	n blocks with 2 layers each i.e. 2n layers
Stack 3	n blocks with 2 layers each i.e. 2n layers
Final Layer	Average pooling + Linear Layer

2. The model has been trained with the following hyperparameters:

- batch_size = 128
- epochs = 100
- momentum = 0.9
- weight_decay = 0.0001
- SGD optimizer with initial learning rate = 0.1, multiplied by 10 at each milestone implemented with the optim.lr_scheduler having gamma = 0.1.

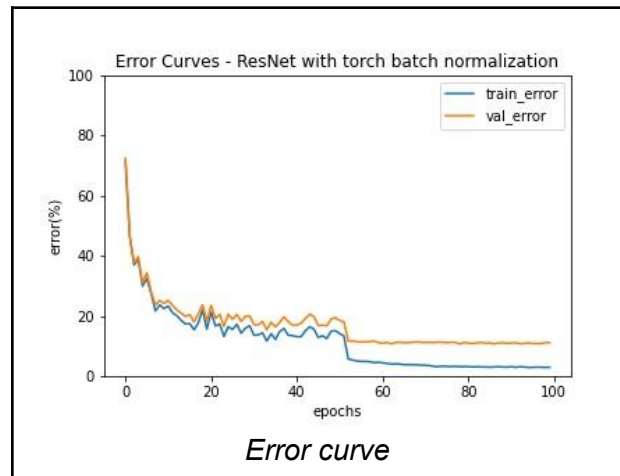
3. Statistics/ Analysis:

- The metrics have been calculated using sklearn.metrics.
- The model with inbuilt batch normalization performs with 96%,96%,89% accuracy on training, validation , and test data respectively.

	precision	recall	f1-score	support		precision	recall	f1-score	support
plane	0.97	0.97	0.97	3963	plane	0.96	0.96	0.96	1037
car	0.99	0.98	0.98	4017	car	0.99	0.98	0.98	983
bird	0.95	0.94	0.95	4019	bird	0.95	0.95	0.95	981
cat	0.92	0.90	0.91	4015	cat	0.90	0.90	0.90	985
deer	0.96	0.95	0.95	4071	deer	0.97	0.95	0.96	929
dog	0.92	0.93	0.93	3975	dog	0.91	0.93	0.92	1025
frog	0.96	0.97	0.97	3964	frog	0.96	0.97	0.97	1036
horse	0.97	0.97	0.97	3968	horse	0.98	0.96	0.97	1032
ship	0.98	0.99	0.98	3979	ship	0.98	0.98	0.98	1021
truck	0.97	0.98	0.98	4029	truck	0.98	0.98	0.98	971
accuracy			0.96	40000	accuracy			0.96	10000
macro avg	0.96	0.96	0.96	40000	macro avg	0.96	0.96	0.96	10000
weighted avg	0.96	0.96	0.96	40000	weighted avg	0.96	0.96	0.96	10000
<i>Metrics on Training Data</i>					<i>Metrics on Validation Data</i>				

	precision	recall	f1-score	support
plane	0.89	0.90	0.89	1000
car	0.95	0.95	0.95	1000
bird	0.85	0.85	0.85	1000
cat	0.80	0.76	0.78	1000
deer	0.88	0.89	0.88	1000
dog	0.84	0.85	0.84	1000
frog	0.91	0.93	0.92	1000
horse	0.93	0.90	0.92	1000
ship	0.93	0.94	0.93	1000
truck	0.92	0.93	0.93	1000
accuracy			0.89	10000
macro avg	0.89	0.89	0.89	10000
weighted avg	0.89	0.89	0.89	10000
<i>Metrics on Test Data</i>				

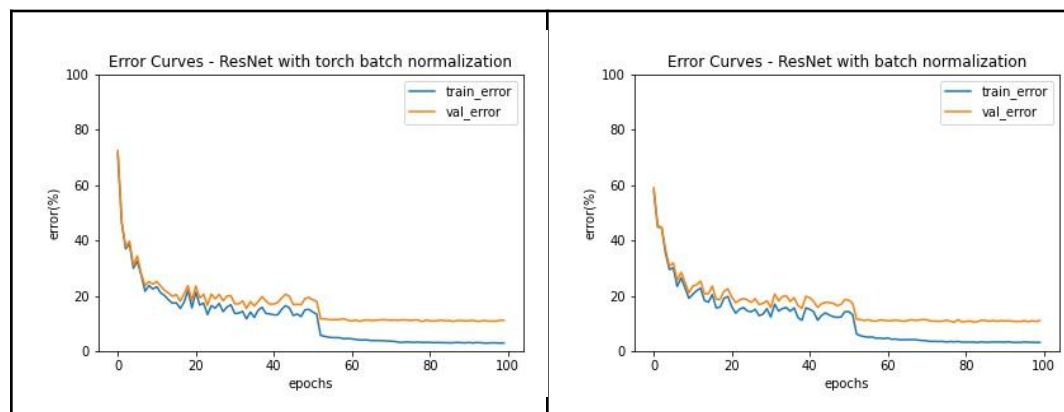
- Error Curves: The error curve shows an expected gradual decrease in training and validation error. Validation error is a bit higher than training.



1.2 - Impact of Normalization

1. Sanity Check with self-implemented BN variant:

- The error curves for the inbuilt and implemented Batch normalization are shown below. The curves are almost similar, thus both the variants are working similarly.



- The metrics for training, validation, and test data are shown below. Both the variants perform similarly on all three datasets.

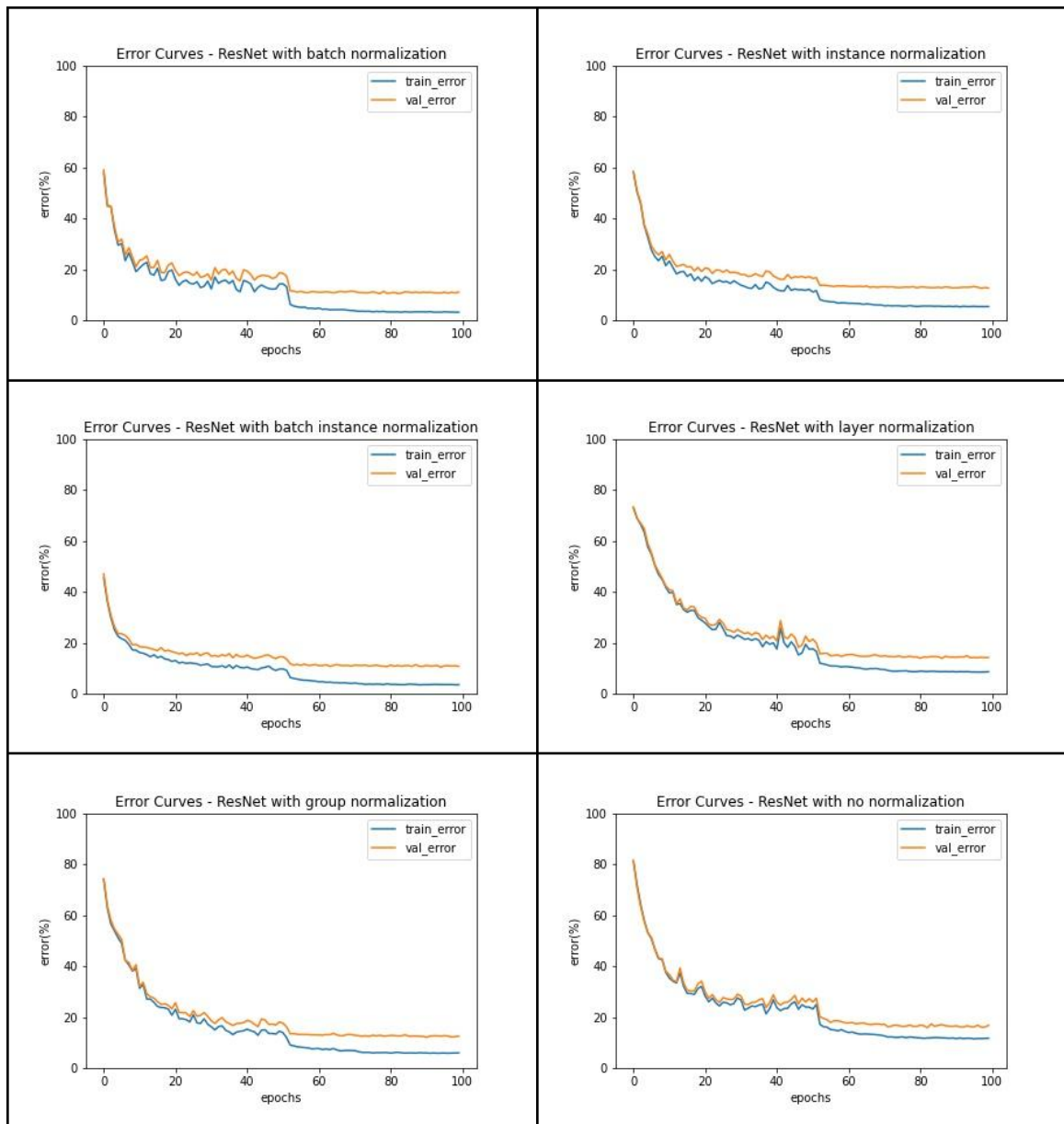
	precision	recall	f1-score	support		precision	recall	f1-score	support
plane	0.97	0.97	0.97	3963	plane	0.96	0.97	0.96	3963
car	0.99	0.98	0.98	4017	car	0.98	0.98	0.98	4017
bird	0.95	0.94	0.95	4019	bird	0.94	0.95	0.94	4019
cat	0.92	0.90	0.91	4015	cat	0.92	0.89	0.90	4015
deer	0.96	0.95	0.95	4071	deer	0.95	0.96	0.96	4071
dog	0.92	0.93	0.93	3975	dog	0.91	0.93	0.92	3975
frog	0.96	0.97	0.97	3964	frog	0.96	0.97	0.97	3964
horse	0.97	0.97	0.97	3968	horse	0.97	0.97	0.97	3968
ship	0.98	0.99	0.98	3979	ship	0.98	0.98	0.98	3979
truck	0.97	0.98	0.98	4029	truck	0.98	0.98	0.98	4029
accuracy			0.96	40000	accuracy			0.96	40000
macro avg	0.96	0.96	0.96	40000	macro avg	0.96	0.96	0.96	40000
weighted avg	0.96	0.96	0.96	40000	weighted avg	0.96	0.96	0.96	40000
Metrics on Training Data					Metrics on Training Data				

	precision	recall	f1-score	support		precision	recall	f1-score	support
plane	0.96	0.96	0.96	1037	plane	0.96	0.97	0.97	1037
car	0.99	0.98	0.98	983	car	0.99	0.98	0.99	983
bird	0.95	0.95	0.95	981	bird	0.94	0.94	0.94	981
cat	0.90	0.90	0.90	985	cat	0.91	0.89	0.90	985
deer	0.97	0.95	0.96	929	deer	0.96	0.96	0.96	929
dog	0.91	0.93	0.92	1025	dog	0.91	0.95	0.93	1025
frog	0.96	0.97	0.97	1036	frog	0.96	0.97	0.97	1036
horse	0.98	0.96	0.97	1032	horse	0.98	0.96	0.97	1032
ship	0.98	0.98	0.98	1021	ship	0.98	0.98	0.98	1021
truck	0.98	0.98	0.98	971	truck	0.97	0.98	0.98	971
accuracy			0.96	10000	accuracy			0.96	10000
macro avg	0.96	0.96	0.96	10000	macro avg	0.96	0.96	0.96	10000
weighted avg	0.96	0.96	0.96	10000	weighted avg	0.96	0.96	0.96	10000
Metrics on Validation Data					Metrics on Validation Data				

	precision	recall	f1-score	support		precision	recall	f1-score	support
plane	0.89	0.90	0.89	1000	plane	0.89	0.92	0.90	1000
car	0.95	0.95	0.95	1000	car	0.95	0.94	0.94	1000
bird	0.85	0.85	0.85	1000	bird	0.86	0.86	0.86	1000
cat	0.80	0.76	0.78	1000	cat	0.81	0.76	0.78	1000
deer	0.88	0.89	0.88	1000	deer	0.89	0.90	0.89	1000
dog	0.84	0.85	0.84	1000	dog	0.84	0.86	0.85	1000
frog	0.91	0.93	0.92	1000	frog	0.92	0.93	0.92	1000
horse	0.93	0.90	0.92	1000	horse	0.92	0.91	0.92	1000
ship	0.93	0.94	0.93	1000	ship	0.94	0.95	0.94	1000
truck	0.92	0.93	0.93	1000	truck	0.93	0.93	0.93	1000
accuracy			0.89	10000	accuracy			0.90	10000
macro avg	0.89	0.89	0.89	10000	macro avg	0.89	0.90	0.89	10000
weighted avg	0.89	0.89	0.89	10000	weighted avg	0.89	0.90	0.89	10000
Metrics on Test Data					Metrics on Test Data				

2. Comparison of six variants

- The error curves for training, validation, and test data are shown below.



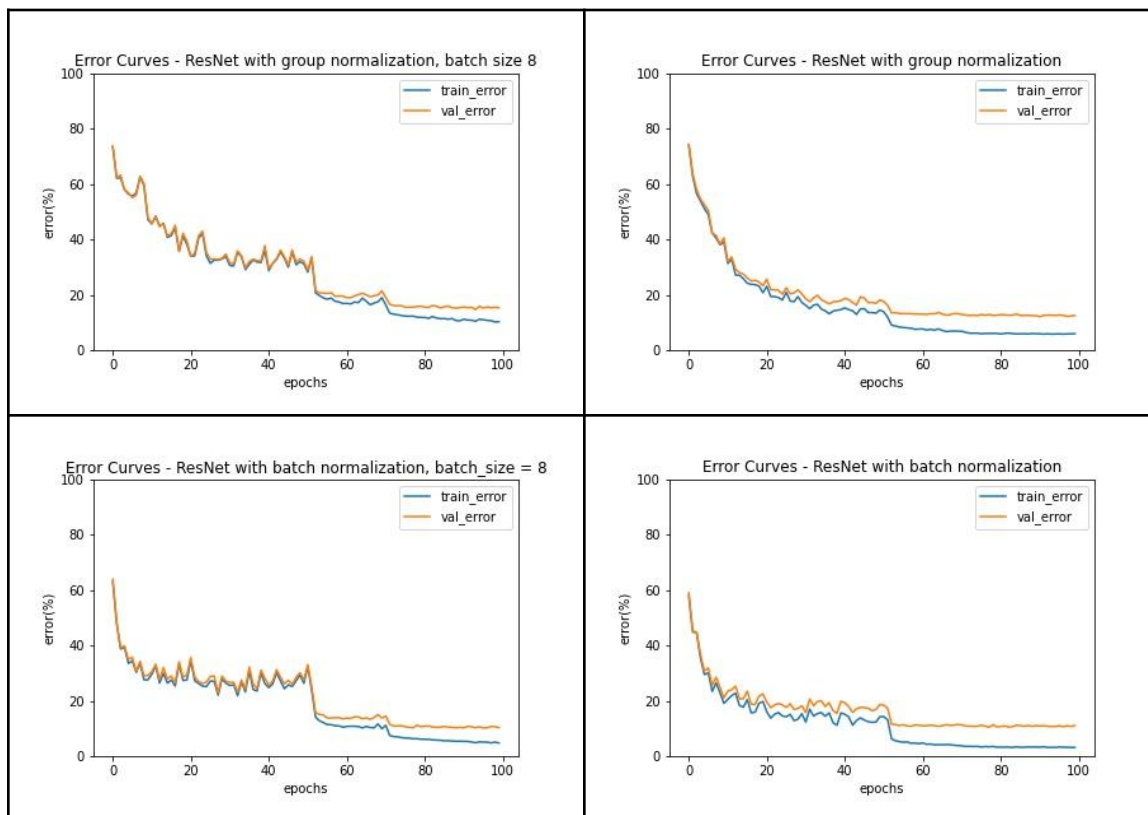
- The metrics for training, validation, and test data are shown below.
- The no normalization variant performs worst among the six variants.

Variants	Training Data	Validation Data	Test Data
Batch normalization	precision recall f1-score support plane 0.96 0.97 0.96 3963 car 0.98 0.98 0.98 4017 bird 0.94 0.95 0.94 4019 cat 0.92 0.89 0.90 4015 deer 0.95 0.96 0.96 4071 dog 0.91 0.93 0.92 3975 frog 0.96 0.97 0.97 3964 horse 0.97 0.97 0.97 3968 ship 0.98 0.98 0.98 3979 truck 0.98 0.98 0.98 4029 accuracy 0.96 40000 macro avg 0.96 0.96 0.96 40000 weighted avg 0.96 0.96 0.96 40000	precision recall f1-score support plane 0.96 0.97 0.97 1037 car 0.99 0.98 0.99 983 bird 0.94 0.94 0.94 981 cat 0.91 0.89 0.90 985 deer 0.96 0.96 0.96 929 dog 0.91 0.95 0.93 1025 frog 0.96 0.97 0.97 1036 horse 0.98 0.96 0.97 1032 ship 0.98 0.98 0.98 1021 truck 0.97 0.98 0.98 971 accuracy 0.96 10000 macro avg 0.96 0.96 0.96 10000 weighted avg 0.96 0.96 0.96 10000	precision recall f1-score support plane 0.89 0.92 0.90 1000 car 0.95 0.94 0.94 1000 bird 0.86 0.86 0.86 1000 cat 0.81 0.76 0.78 1000 deer 0.89 0.90 0.89 1000 dog 0.84 0.86 0.85 1000 frog 0.92 0.93 0.92 1000 horse 0.92 0.91 0.92 1000 ship 0.94 0.95 0.94 1000 truck 0.93 0.93 0.93 1000 accuracy 0.90 10000 macro avg 0.89 0.90 0.89 10000 weighted avg 0.89 0.90 0.89 10000
Instance Normalization	precision recall f1-score support plane 0.94 0.93 0.94 3973 car 0.97 0.97 0.97 4067 bird 0.93 0.91 0.92 4006 cat 0.86 0.88 0.87 4044 deer 0.95 0.93 0.94 3962 dog 0.90 0.89 0.89 3937 frog 0.94 0.96 0.95 4032 horse 0.95 0.96 0.95 3978 ship 0.97 0.97 0.97 4036 truck 0.96 0.97 0.96 3965 accuracy 0.94 40000 macro avg 0.94 0.94 0.94 40000 weighted avg 0.94 0.94 0.94 40000	precision recall f1-score support plane 0.96 0.93 0.94 1027 car 0.97 0.97 0.97 933 bird 0.93 0.91 0.92 994 cat 0.84 0.88 0.86 956 deer 0.94 0.92 0.93 1038 dog 0.90 0.88 0.89 1063 frog 0.94 0.96 0.95 968 horse 0.93 0.96 0.95 1022 ship 0.95 0.97 0.96 964 truck 0.97 0.97 0.97 1035 accuracy 0.93 10000 macro avg 0.93 0.93 0.93 10000 weighted avg 0.93 0.93 0.93 10000	precision recall f1-score support plane 0.87 0.87 0.87 1000 car 0.95 0.94 0.94 1000 bird 0.85 0.81 0.83 1000 cat 0.75 0.74 0.75 1000 deer 0.88 0.87 0.87 1000 dog 0.82 0.82 0.82 1000 frog 0.88 0.91 0.89 1000 horse 0.91 0.92 0.91 1000 ship 0.91 0.92 0.92 1000 truck 0.90 0.93 0.92 1000 accuracy 0.87 10000 macro avg 0.87 0.87 0.87 10000 weighted avg 0.87 0.87 0.87 10000
Batch-Instance Normalization	precision recall f1-score support plane 0.93 0.93 0.93 4025 car 0.97 0.96 0.96 4017 bird 0.92 0.91 0.91 3984 cat 0.84 0.89 0.87 4005 deer 0.93 0.93 0.93 3991 dog 0.90 0.87 0.89 4022 frog 0.95 0.95 0.95 3983 horse 0.95 0.95 0.95 3961 ship 0.96 0.96 0.96 4031 truck 0.95 0.95 0.95 3981 accuracy 0.93 40000 macro avg 0.93 0.93 0.93 40000 weighted avg 0.93 0.93 0.93 40000	precision recall f1-score support plane 0.92 0.93 0.93 975 car 0.97 0.96 0.97 983 bird 0.91 0.91 0.91 1016 cat 0.84 0.87 0.86 995 deer 0.92 0.93 0.92 1009 dog 0.90 0.86 0.88 978 frog 0.95 0.94 0.95 1017 horse 0.95 0.95 0.95 1039 ship 0.95 0.96 0.95 969 truck 0.96 0.95 0.96 1019 accuracy 0.93 10000 macro avg 0.93 0.93 0.93 10000 weighted avg 0.93 0.93 0.93 10000	precision recall f1-score support plane 0.92 0.93 0.93 975 car 0.97 0.96 0.97 983 bird 0.91 0.91 0.91 1016 cat 0.84 0.87 0.86 995 deer 0.92 0.93 0.92 1009 dog 0.90 0.86 0.88 978 frog 0.95 0.94 0.95 1017 horse 0.95 0.95 0.95 1039 ship 0.95 0.96 0.95 969 truck 0.96 0.95 0.96 1019 accuracy 0.93 10000 macro avg 0.93 0.93 0.93 10000 weighted avg 0.93 0.93 0.93 10000
Layer Normalization	precision recall f1-score support plane 0.91 0.91 0.91 3979 car 0.96 0.95 0.95 4027 bird 0.88 0.86 0.87 4036 cat 0.78 0.78 0.78 4018 deer 0.89 0.90 0.89 4029 dog 0.81 0.82 0.81 3957 frog 0.93 0.92 0.92 3997 horse 0.91 0.92 0.92 3989 ship 0.95 0.95 0.95 3989 truck 0.92 0.94 0.93 3979 accuracy 0.89 40000 macro avg 0.89 0.89 0.89 40000 weighted avg 0.89 0.89 0.89 40000	precision recall f1-score support plane 0.91 0.93 0.92 1021 car 0.96 0.95 0.96 973 bird 0.87 0.86 0.87 964 cat 0.77 0.79 0.78 982 deer 0.90 0.91 0.90 971 dog 0.83 0.84 0.83 1043 frog 0.93 0.90 0.92 1003 horse 0.92 0.91 0.91 1011 ship 0.95 0.95 0.95 1011 truck 0.94 0.95 0.94 1021 accuracy 0.90 10000 macro avg 0.90 0.90 0.90 10000 weighted avg 0.90 0.90 0.90 10000	precision recall f1-score support plane 0.83 0.87 0.85 1000 car 0.94 0.92 0.93 1000 bird 0.80 0.78 0.79 1000 cat 0.72 0.69 0.70 1000 deer 0.82 0.86 0.84 1000 dog 0.77 0.77 0.77 1000 frog 0.89 0.88 0.88 1000 horse 0.88 0.85 0.87 1000 ship 0.92 0.91 0.91 1000 truck 0.87 0.92 0.89 1000 accuracy 0.84 10000 macro avg 0.84 0.84 0.84 10000 weighted avg 0.84 0.84 0.84 10000

Group Normalization	precision recall f1-score support				precision recall f1-score support				precision recall f1-score support						
	plane	0.93	0.94	0.94	3973	plane	0.94	0.94	0.94	1027	plane	0.87	0.87	0.87	1000
	car	0.97	0.97	0.97	4067	car	0.97	0.98	0.97	933	car	0.93	0.95	0.94	1000
	bird	0.91	0.91	0.91	4006	bird	0.92	0.90	0.91	994	bird	0.84	0.84	0.84	1000
	cat	0.84	0.86	0.85	4044	cat	0.83	0.87	0.85	956	cat	0.76	0.77	0.77	1000
	deer	0.95	0.92	0.93	3962	deer	0.94	0.93	0.93	1038	deer	0.86	0.87	0.87	1000
	dog	0.89	0.87	0.88	3937	dog	0.90	0.87	0.89	1063	dog	0.83	0.80	0.82	1000
	frog	0.95	0.95	0.95	4032	frog	0.94	0.94	0.94	968	frog	0.91	0.90	0.90	1000
	horse	0.95	0.95	0.95	3978	horse	0.94	0.94	0.94	1022	horse	0.91	0.90	0.91	1000
	ship	0.97	0.97	0.97	4036	ship	0.97	0.97	0.97	964	ship	0.94	0.91	0.93	1000
truck	0.95	0.97	0.96	3965	truck	0.96	0.97	0.96	1035	truck	0.90	0.93	0.91	1000	
	accuracy			0.93	40000	accuracy			0.93	10000	accuracy			0.88	10000
	macro avg	0.93	0.93	0.93	40000	macro avg	0.93	0.93	0.93	10000	macro avg	0.88	0.88	0.88	10000
	weighted avg	0.93	0.93	0.93	40000	weighted avg	0.93	0.93	0.93	10000	weighted avg	0.88	0.88	0.88	10000
No Normalization	precision recall f1-score support				precision recall f1-score support				precision recall f1-score support						
	plane	0.88	0.91	0.90	3973	plane	0.90	0.92	0.91	1027	plane	0.81	0.88	0.84	1000
	car	0.96	0.95	0.95	4067	car	0.97	0.95	0.96	933	car	0.93	0.92	0.92	1000
	bird	0.85	0.83	0.84	4006	bird	0.85	0.84	0.84	994	bird	0.79	0.76	0.77	1000
	cat	0.76	0.75	0.76	4044	cat	0.74	0.75	0.74	956	cat	0.70	0.66	0.68	1000
	deer	0.88	0.86	0.87	3962	deer	0.88	0.87	0.88	1038	deer	0.84	0.81	0.83	1000
	dog	0.79	0.81	0.80	3937	dog	0.81	0.79	0.80	1063	dog	0.72	0.76	0.74	1000
	frog	0.91	0.90	0.91	4032	frog	0.90	0.90	0.90	968	frog	0.87	0.87	0.87	1000
	horse	0.91	0.91	0.91	3978	horse	0.90	0.91	0.91	1022	horse	0.89	0.86	0.88	1000
	ship	0.94	0.95	0.95	4036	ship	0.94	0.96	0.95	964	ship	0.90	0.91	0.91	1000
truck	0.93	0.94	0.94	3965	truck	0.94	0.95	0.95	1035	truck	0.88	0.91	0.89	1000	
	accuracy			0.88	40000	accuracy			0.88	10000	accuracy			0.83	10000
	macro avg	0.88	0.88	0.88	40000	macro avg	0.88	0.88	0.88	10000	macro avg	0.83	0.83	0.83	10000
	weighted avg	0.88	0.88	0.88	40000	weighted avg	0.88	0.88	0.88	10000	weighted avg	0.83	0.83	0.83	10000

3. Impact of Batch Size

- Group normalization has a similar trajectory with slight perturbations, while batch normalization shows large changes.



4. Evolution of feature distributions

- The feature evolutions were calculated from the penultimate layer, i.e., global average pooling layer, having a dimension of 640000 for the 10000 image data from the validation set.

