

Computer Vision HW2 Report

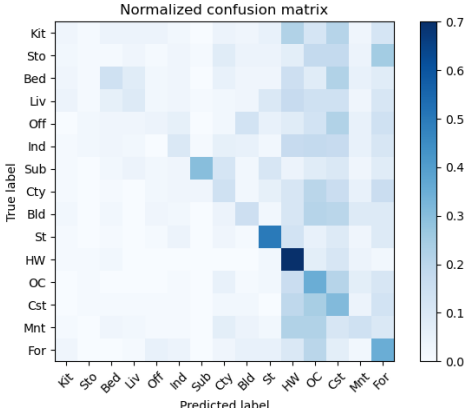
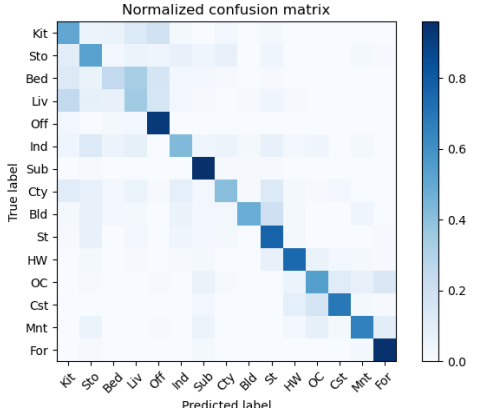
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Part 1. (10%)

- Plot confusion matrix of two settings. (i.e. Bag of sift and tiny image representation) (5%)

Ans:

	Tiny image	Bag of sift
confusion matrix		
k	1	1
accuracy	0.22333333333333333	0.6133333333333333

- Compare the results/accuracy of both settings and explain the result. (5%)

Ans:

As the result shown above, the performance of bag of sift is much better than tiny image. Tiny image is not a particularly good representation, because it discards all of the high frequency image content and is not especially invariant to spatial or brightness shifts.

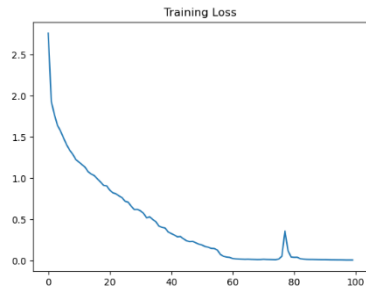
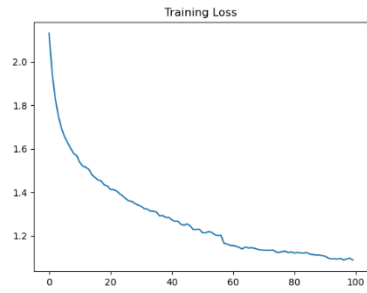
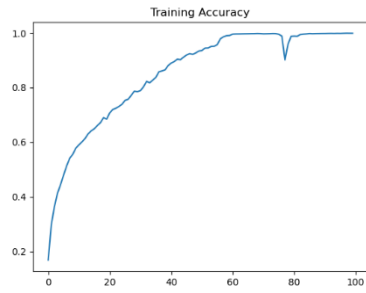
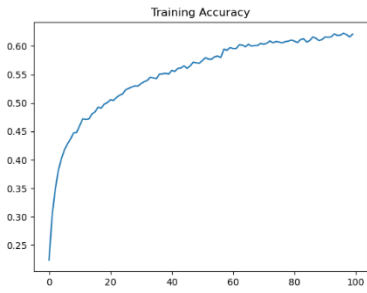
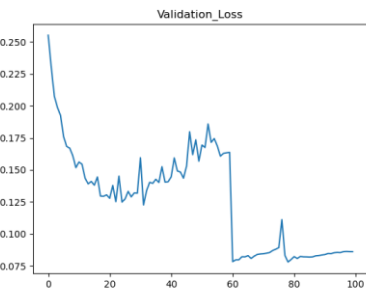
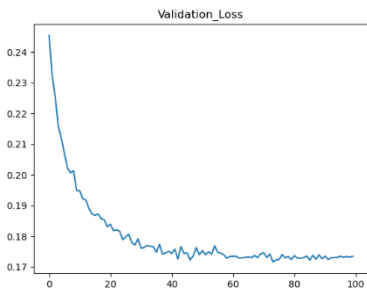
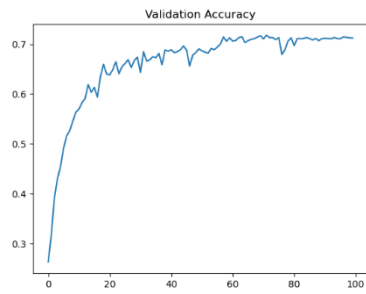
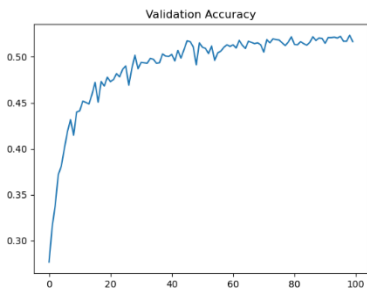
Part 2. (35%)

- Compare the performance on residual networks and LeNet. Plot the learning curve (loss and accuracy) on both training and validation sets for both 2 schemes. 8 plots in total. (20%)

Ans:

From the following plots, we can observed that the residual network converged a bit earlier than LeNet.

	Residual networks	LeNet
Accuracy (public test)	0.814	0.5976

Training	Loss		
	Accuracy		
Validation	Loss		
	Accuracy		

• **Attach basic information of the model you use including model architecture and the number of the parameters. Besides, report the accuracy you performed on the public test set. (5%)**

Ans:

- accuracy on the public test set: 0.823
- model architecture and the number of the parameters: 24,562,250

layer (type:depth-idx)	Param #
Conv2d: 1-1	9,408
BatchNorm2d: 1-2	128
ReLU: 1-3	--
MaxPool2d: 1-4	--
Sequential: 1-5	--
Bottleneck: 2-1	--
Conv2d: 3-1	4,096
BatchNorm2d: 3-2	128
Conv2d: 3-3	36,864
BatchNorm2d: 3-4	128
Conv2d: 3-5	16,384
BatchNorm2d: 3-6	512
ReLU: 3-7	--
Sequential: 3-8	16,896
Bottleneck: 2-2	--
Conv2d: 3-9	16,384
BatchNorm2d: 3-10	128
Conv2d: 3-11	36,864
BatchNorm2d: 3-12	128
Conv2d: 3-13	16,384
BatchNorm2d: 3-14	512
ReLU: 3-15	--
Bottleneck: 2-3	--
Conv2d: 3-16	16,384
BatchNorm2d: 3-17	128
Conv2d: 3-18	36,864
BatchNorm2d: 3-19	128
Conv2d: 3-20	16,384
BatchNorm2d: 3-21	512
ReLU: 3-22	--
Sequential: 1-6	--
Bottleneck: 2-4	--
Conv2d: 3-23	32,768
BatchNorm2d: 3-24	256
Conv2d: 3-25	147,456
BatchNorm2d: 3-26	256
Conv2d: 3-27	65,536
BatchNorm2d: 3-28	1,024
ReLU: 3-29	--
Sequential: 3-30	132,096
Bottleneck: 2-5	--
Conv2d: 3-31	65,536
BatchNorm2d: 3-32	256
Conv2d: 3-33	147,456
BatchNorm2d: 3-34	256
Conv2d: 3-35	65,536
BatchNorm2d: 3-36	1,024
ReLU: 3-37	--
Bottleneck: 2-6	--
Conv2d: 3-38	65,536
BatchNorm2d: 3-39	256
Conv2d: 3-40	147,456
BatchNorm2d: 3-41	256
Conv2d: 3-42	65,536
BatchNorm2d: 3-43	1,024
ReLU: 3-44	--
Bottleneck: 2-7	--
Conv2d: 3-45	65,536
BatchNorm2d: 3-46	256
Conv2d: 3-47	147,456
BatchNorm2d: 3-48	256
Conv2d: 3-49	65,536
BatchNorm2d: 3-50	1,024
ReLU: 3-51	--
Sequential: 1-7	--
Bottleneck: 2-8	--
Conv2d: 3-52	131,072
BatchNorm2d: 3-53	512
Conv2d: 3-54	589,824
BatchNorm2d: 3-55	512
Conv2d: 3-56	262,144
BatchNorm2d: 3-57	2,048
ReLU: 3-58	--
Sequential: 3-59	526,336
Bottleneck: 2-9	--
Conv2d: 3-60	262,144
BatchNorm2d: 3-61	512
Conv2d: 3-62	589,824
BatchNorm2d: 3-63	512
Conv2d: 3-64	262,144
BatchNorm2d: 3-65	2,048
ReLU: 3-66	--
Bottleneck: 2-10	--
Conv2d: 3-67	262,144
BatchNorm2d: 3-68	512
Conv2d: 3-69	589,824
BatchNorm2d: 3-70	512
Conv2d: 3-71	262,144
BatchNorm2d: 3-72	2,048
ReLU: 3-73	--
Bottleneck: 2-11	--
Conv2d: 3-74	262,144
BatchNorm2d: 3-75	512
Conv2d: 3-76	589,824
BatchNorm2d: 3-77	512
Conv2d: 3-78	262,144
BatchNorm2d: 3-79	2,048
ReLU: 3-80	--
Bottleneck: 2-12	--
Conv2d: 3-81	262,144
BatchNorm2d: 3-82	512
Conv2d: 3-83	589,824
BatchNorm2d: 3-84	512
Conv2d: 3-85	262,144
BatchNorm2d: 3-86	2,048
ReLU: 3-87	--
Bottleneck: 2-13	--
Conv2d: 3-88	262,144
BatchNorm2d: 3-89	512
Conv2d: 3-90	589,824
BatchNorm2d: 3-91	512
Conv2d: 3-92	262,144
BatchNorm2d: 3-93	2,048
ReLU: 3-94	--
Sequential: 1-8	--
Bottleneck: 2-14	--
Conv2d: 3-95	524,288
BatchNorm2d: 3-96	1,024
Conv2d: 3-97	2,359,296
BatchNorm2d: 3-98	1,024
Conv2d: 3-99	1,048,576
BatchNorm2d: 3-100	4,096
ReLU: 3-101	--
Sequential: 3-102	2,101,248
Bottleneck: 2-15	--
Conv2d: 3-103	1,048,576
BatchNorm2d: 3-104	1,024
Conv2d: 3-105	2,359,296
BatchNorm2d: 3-106	1,024
Conv2d: 3-107	1,048,576
BatchNorm2d: 3-108	4,096
ReLU: 3-109	--
Bottleneck: 2-16	--
Conv2d: 3-110	1,048,576
BatchNorm2d: 3-111	1,024
Conv2d: 3-112	2,359,296
BatchNorm2d: 3-113	1,024
Conv2d: 3-114	1,048,576
BatchNorm2d: 3-115	4,096
ReLU: 3-116	--
AdaptiveAvgPool2d: 1-9	--
Sequential: 1-10	--
Linear: 2-17	1,049,088
ReLU: 2-18	--
Dropout: 2-19	--
Linear: 2-20	5,130
Total params: 24,562,250	
Trainable params: 24,562,250	
Non-trainable params: 0	

• Briefly describe what method do you apply? (e.g. data augmentation, model architecture, loss function, semi-supervised etc.) (10%)

Ans:

- data augmentation: RandomHorizontalFlip(p=0.5), RandomRotation(degrees=20),
- model architecture: add a dropout layer in fully-connected layers to prevent model from overfitting
- loss function: CrossEntropyLoss()
- semi-supervised: After the accuracy on validation set reached 0.7, I used the model at that point to generate pseudo labels. The images and the corresponding pseudo labels whose confidence was greater than 0.9 were added to the training set. This method gave me approximately 0.012 improvement on accuracy.