

# Assignment 2 report: The German Traffic Sign Recognition Benchmark — Rev 1

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**Honor Statement:** By submitting this work, I certify that, with the exception of LaTeX templates and text whose sources I cite, every keystroke in the [answers](#) was typed by me.

Instructor: Prof. Rob Fergus.

## 1 The environment for this assignment

- OS: Windows 10 x64
- CUDA: 10.0
- Python: 3.7.3
- CUDNN: 7
- pytorch: 1.2.0
- GPU: GeForce GTX 1070 (8+8 GB GPU Memory)
- CPU: i7-8750H

P.S. Estimated time to run one batch on this PC: 5 minutes

## 2 About the dataset

1. The shape of the sample:  $[*, 32, 32]$ , which means the sample is RGB with 32 x 32 pixel
2. The training dataset has 35339 samples and the validation dataset has 3870 samples.

## 3 My framework and the hyper-parameters

I used ensemble models which includes 4 GoogLeNets and a Basic CNN Net. 3 GoogLeNets are trained in the same process (Please see the GoogLeNet-Test-3.ipynb) and the another is trained in another process (Please see the GoogLeNet-Test-2.ipynb). I used the same random seed for the two process of training GoogLeNets. However, the result of the two process are actually different.

How do I choose those GoogLeNets?

In GoogLeNet-Test-3.ipynb, by looking at the log and choose those accuracy are local maximum which means not only that epoch has a big validation accuracy but also the near epochs have relatively larger validation accuracy.

1. GoogLeNet
2. A Basic CNN Net

I will show the frameworks at the end of this file.

## **4 How I ensemble these models?**

During the process of get the Kaggle files, I got the raw classification scores for each model and take the mean. Then, I use it as the ensemble model result which performs better than any models in the ensemble list.

Please also see the training loss graph and accuracy graph in the end of this doc

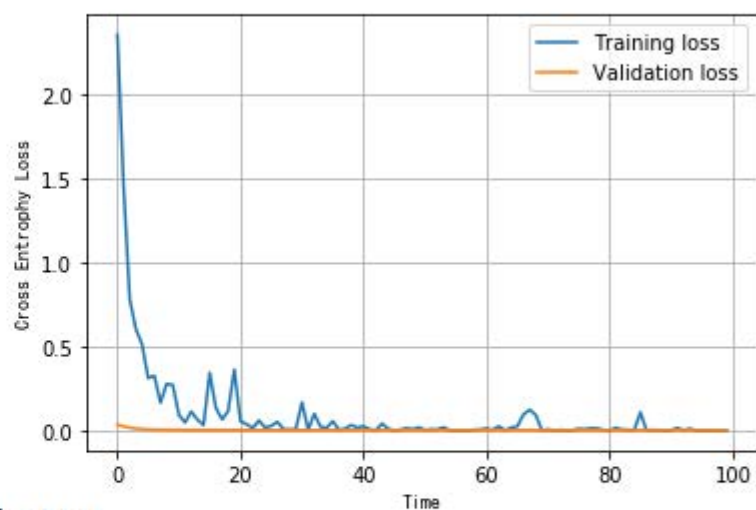
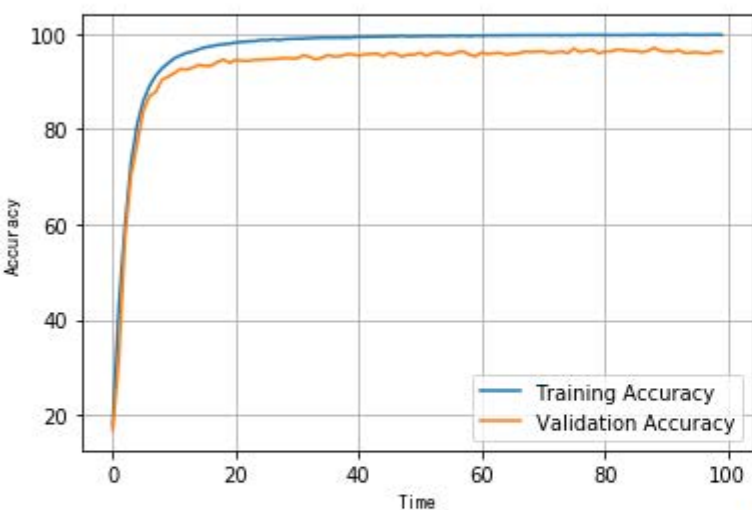
Layer (type)	Output Shape	Param #	Layer (type)	Output Shape	Param #
Conv2d-1	[-1, 192, 32, 32]	5,376	ReLU-108	[-1, 24, 16, 16]	0
BatchNorm2d-2	[-1, 192, 32, 32]	384	Conv2d-109	[-1, 64, 16, 16]	13,488
ReLU-3	[-1, 192, 32, 32]	0	BatchNorm2d-110	[-1, 64, 16, 16]	128
Conv2d-4	[-1, 64, 32, 32]	12,352	ReLU-111	[-1, 64, 16, 16]	0
BatchNorm2d-5	[-1, 64, 32, 32]	128	Conv2d-112	[-1, 64, 16, 16]	36,928
ReLU-6	[-1, 64, 32, 32]	0	BatchNorm2d-113	[-1, 64, 16, 16]	128
Conv2d-7	[-1, 96, 32, 32]	18,528	ReLU-114	[-1, 64, 16, 16]	0
BatchNorm2d-8	[-1, 96, 32, 32]	192	MaxPool2d-115	[-1, 512, 16, 16]	0
ReLU-9	[-1, 96, 32, 32]	0	Conv2d-116	[-1, 64, 16, 16]	32,832
Conv2d-10	[-1, 128, 32, 32]	110,720	BatchNorm2d-117	[-1, 64, 16, 16]	128
BatchNorm2d-11	[-1, 128, 32, 32]	256	ReLU-118	[-1, 64, 16, 16]	0
ReLU-12	[-1, 128, 32, 32]	0	Inception-119	[-1, 512, 16, 16]	0
Conv2d-13	[-1, 16, 32, 32]	3,088	Conv2d-120	[-1, 112, 16, 16]	57,456
BatchNorm2d-14	[-1, 16, 32, 32]	32	BatchNorm2d-121	[-1, 112, 16, 16]	224
ReLU-15	[-1, 16, 32, 32]	0	ReLU-122	[-1, 112, 16, 16]	0
Conv2d-16	[-1, 32, 32, 32]	4,640	Conv2d-123	[-1, 144, 16, 16]	73,872
BatchNorm2d-17	[-1, 32, 32, 32]	64	BatchNorm2d-124	[-1, 144, 16, 16]	288
ReLU-18	[-1, 32, 32, 32]	0	ReLU-125	[-1, 144, 16, 16]	0
Conv2d-19	[-1, 32, 32, 32]	9,248	Conv2d-126	[-1, 288, 16, 16]	373,536
BatchNorm2d-20	[-1, 32, 32, 32]	64	BatchNorm2d-127	[-1, 288, 16, 16]	576
ReLU-21	[-1, 32, 32, 32]	0	ReLU-128	[-1, 288, 16, 16]	0
MaxPool2d-22	[-1, 192, 32, 32]	0	Conv2d-129	[-1, 32, 16, 16]	16,416
Conv2d-23	[-1, 32, 32, 32]	6,176	BatchNorm2d-130	[-1, 32, 16, 16]	64
BatchNorm2d-24	[-1, 32, 32, 32]	64	ReLU-131	[-1, 32, 16, 16]	0
ReLU-25	[-1, 32, 32, 32]	0	Conv2d-132	[-1, 64, 16, 16]	18,496
Inception-26	[-1, 256, 32, 32]	0	BatchNorm2d-133	[-1, 64, 16, 16]	128
Conv2d-27	[-1, 128, 32, 32]	32,896	ReLU-134	[-1, 64, 16, 16]	0
BatchNorm2d-28	[-1, 128, 32, 32]	256	Conv2d-135	[-1, 64, 16, 16]	36,928
ReLU-29	[-1, 128, 32, 32]	0	BatchNorm2d-136	[-1, 64, 16, 16]	128
Conv2d-30	[-1, 128, 32, 32]	32,896	ReLU-137	[-1, 64, 16, 16]	0
BatchNorm2d-31	[-1, 128, 32, 32]	256	MaxPool2d-138	[-1, 512, 16, 16]	0
ReLU-32	[-1, 128, 32, 32]	0	Conv2d-139	[-1, 64, 16, 16]	32,832
Conv2d-33	[-1, 192, 32, 32]	221,376	BatchNorm2d-140	[-1, 64, 16, 16]	128
BatchNorm2d-34	[-1, 192, 32, 32]	384	ReLU-141	[-1, 64, 16, 16]	0
ReLU-35	[-1, 192, 32, 32]	0	Inception-142	[-1, 528, 16, 16]	0
Conv2d-36	[-1, 32, 32, 32]	8,224	Conv2d-143	[-1, 256, 16, 16]	135,424
BatchNorm2d-37	[-1, 32, 32, 32]	64	BatchNorm2d-144	[-1, 256, 16, 16]	512
ReLU-38	[-1, 32, 32, 32]	0	ReLU-145	[-1, 256, 16, 16]	0
Conv2d-39	[-1, 96, 32, 32]	27,744	Conv2d-146	[-1, 160, 16, 16]	84,640
BatchNorm2d-40	[-1, 96, 32, 32]	192	BatchNorm2d-147	[-1, 160, 16, 16]	320
ReLU-41	[-1, 96, 32, 32]	0	ReLU-148	[-1, 160, 16, 16]	0
Conv2d-42	[-1, 96, 32, 32]	83,040	Conv2d-149	[-1, 320, 16, 16]	461,120
BatchNorm2d-43	[-1, 96, 32, 32]	192	BatchNorm2d-150	[-1, 320, 16, 16]	640
ReLU-44	[-1, 96, 32, 32]	0	ReLU-151	[-1, 320, 16, 16]	0
MaxPool2d-45	[-1, 256, 32, 32]	0	Conv2d-152	[-1, 32, 16, 16]	16,928
Conv2d-46	[-1, 64, 32, 32]	16,448	BatchNorm2d-153	[-1, 32, 16, 16]	64
BatchNorm2d-47	[-1, 64, 32, 32]	128	ReLU-154	[-1, 32, 16, 16]	0
ReLU-48	[-1, 64, 32, 32]	0	Conv2d-155	[-1, 128, 16, 16]	36,992
Inception-49	[-1, 480, 32, 32]	0	BatchNorm2d-156	[-1, 128, 16, 16]	256
MaxPool2d-50	[-1, 480, 16, 16]	0	ReLU-157	[-1, 128, 16, 16]	0
Conv2d-51	[-1, 192, 16, 16]	92,352	Conv2d-158	[-1, 128, 16, 16]	147,584
BatchNorm2d-52	[-1, 192, 16, 16]	384	BatchNorm2d-159	[-1, 128, 16, 16]	256
ReLU-53	[-1, 192, 16, 16]	0	ReLU-160	[-1, 128, 16, 16]	0
Conv2d-54	[-1, 96, 16, 16]	46,176	MaxPool2d-161	[-1, 528, 16, 16]	0
BatchNorm2d-55	[-1, 96, 16, 16]	192	Conv2d-162	[-1, 128, 16, 16]	67,712
ReLU-56	[-1, 96, 16, 16]	0	BatchNorm2d-163	[-1, 128, 16, 16]	256
Conv2d-57	[-1, 288, 16, 16]	179,920	ReLU-164	[-1, 128, 16, 16]	0
BatchNorm2d-58	[-1, 288, 16, 16]	416	Inception-165	[-1, 832, 16, 16]	0
ReLU-59	[-1, 288, 16, 16]	0	MaxPool2d-166	[-1, 832, 8, 8]	0
Conv2d-60	[-1, 16, 16, 16]	7,696	Conv2d-167	[-1, 256, 8, 8]	213,248
BatchNorm2d-61	[-1, 16, 16, 16]	32	BatchNorm2d-168	[-1, 256, 8, 8]	512
ReLU-62	[-1, 16, 16, 16]	0	ReLU-169	[-1, 256, 8, 8]	0
Conv2d-63	[-1, 48, 16, 16]	6,960	Conv2d-170	[-1, 160, 8, 8]	133,280
BatchNorm2d-64	[-1, 48, 16, 16]	96	BatchNorm2d-171	[-1, 160, 8, 8]	320
ReLU-65	[-1, 48, 16, 16]	0	ReLU-172	[-1, 160, 8, 8]	0
Conv2d-66	[-1, 48, 16, 16]	20,784	Conv2d-173	[-1, 320, 8, 8]	461,120
BatchNorm2d-67	[-1, 48, 16, 16]	96	BatchNorm2d-174	[-1, 320, 8, 8]	640
ReLU-68	[-1, 48, 16, 16]	0	ReLU-175	[-1, 320, 8, 8]	0
MaxPool2d-69	[-1, 480, 16, 16]	0	Conv2d-176	[-1, 32, 8, 8]	26,656
Conv2d-70	[-1, 64, 16, 16]	30,784	BatchNorm2d-177	[-1, 32, 8, 8]	64
BatchNorm2d-71	[-1, 64, 16, 16]	128	ReLU-178	[-1, 32, 8, 8]	0
ReLU-72	[-1, 64, 16, 16]	0	Conv2d-179	[-1, 128, 8, 8]	36,992
Inception-73	[-1, 512, 16, 16]	0	BatchNorm2d-180	[-1, 128, 8, 8]	256
Conv2d-74	[-1, 160, 16, 16]	82,080	ReLU-181	[-1, 128, 8, 8]	0
BatchNorm2d-75	[-1, 160, 16, 16]	320	Conv2d-182	[-1, 128, 8, 8]	147,584
ReLU-76	[-1, 160, 16, 16]	0	BatchNorm2d-183	[-1, 128, 8, 8]	256
Conv2d-77	[-1, 112, 16, 16]	57,456	ReLU-184	[-1, 128, 8, 8]	0
BatchNorm2d-78	[-1, 112, 16, 16]	224	MaxPool2d-185	[-1, 832, 8, 8]	0
ReLU-79	[-1, 112, 16, 16]	0	Conv2d-186	[-1, 128, 8, 8]	106,624
Conv2d-80	[-1, 224, 16, 16]	226,016	BatchNorm2d-187	[-1, 128, 8, 8]	256
BatchNorm2d-81	[-1, 224, 16, 16]	448	ReLU-188	[-1, 128, 8, 8]	0
ReLU-82	[-1, 224, 16, 16]	0	Inception-189	[-1, 832, 8, 8]	0
Conv2d-83	[-1, 24, 16, 16]	12,312	Conv2d-190	[-1, 384, 8, 8]	319,872
BatchNorm2d-84	[-1, 24, 16, 16]	48	BatchNorm2d-191	[-1, 384, 8, 8]	768
ReLU-85	[-1, 24, 16, 16]	0	ReLU-192	[-1, 384, 8, 8]	0
Conv2d-86	[-1, 64, 16, 16]	13,888	Conv2d-193	[-1, 192, 8, 8]	159,936
BatchNorm2d-87	[-1, 64, 16, 16]	128	BatchNorm2d-194	[-1, 192, 8, 8]	384
ReLU-88	[-1, 64, 16, 16]	0	ReLU-195	[-1, 192, 8, 8]	0
Conv2d-89	[-1, 64, 16, 16]	36,928	Conv2d-196	[-1, 384, 8, 8]	663,936
BatchNorm2d-90	[-1, 64, 16, 16]	128	BatchNorm2d-197	[-1, 384, 8, 8]	768
ReLU-91	[-1, 64, 16, 16]	0	ReLU-198	[-1, 384, 8, 8]	0
MaxPool2d-92	[-1, 512, 16, 16]	0	Conv2d-199	[-1, 48, 8, 8]	39,984
Conv2d-93	[-1, 64, 16, 16]	32,832	BatchNorm2d-200	[-1, 48, 8, 8]	96
BatchNorm2d-94	[-1, 64, 16, 16]	128	ReLU-201	[-1, 48, 8, 8]	0
ReLU-95	[-1, 64, 16, 16]	0	Conv2d-202	[-1, 128, 8, 8]	55,424
Inception-96	[-1, 128, 16, 16]	0	BatchNorm2d-203	[-1, 128, 8, 8]	256
Conv2d-97	[-1, 128, 16, 16]	65,664	ReLU-204	[-1, 128, 8, 8]	0
BatchNorm2d-98	[-1, 128, 16, 16]	256	Conv2d-205	[-1, 128, 8, 8]	147,584
ReLU-99	[-1, 128, 16, 16]	0	BatchNorm2d-206	[-1, 128, 8, 8]	256
Conv2d-100	[-1, 128, 16, 16]	65,664	ReLU-207	[-1, 128, 8, 8]	0
BatchNorm2d-101	[-1, 128, 16, 16]	256	MaxPool2d-208	[-1, 832, 8, 8]	0
ReLU-102	[-1, 128, 16, 16]	0	Conv2d-209	[-1, 128, 8, 8]	106,624
Conv2d-103	[-1, 256, 16, 16]	295,168	BatchNorm2d-210	[-1, 128, 8, 8]	256
BatchNorm2d-104	[-1, 256, 16, 16]	512	ReLU-211	[-1, 128, 8, 8]	0
ReLU-105	[-1, 256, 16, 16]	0	Inception-212	[-1, 1024, 8, 8]	0
Conv2d-106	[-1, 24, 16, 16]	12,312	AvgPool2d-213	[-1, 1024, 1, 1]	0
BatchNorm2d-107	[-1, 24, 16, 16]	48	Dropout2d-214	[-1, 1024, 1, 1]	0
			Linear-215	[-1, 43]	44,075

Total params: 6,200,075  
Trainable params: 6,200,075  
Non-trainable params: 0  
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Input size (MB): 0.01  
Forward/backward pass size (MB): 81.43  
Params size (MB): 23.65  
Estimated Total Size (MB): 105.09  
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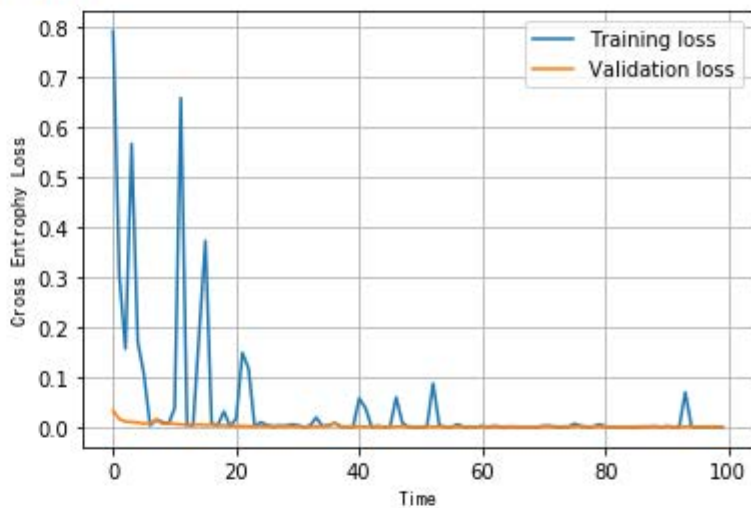
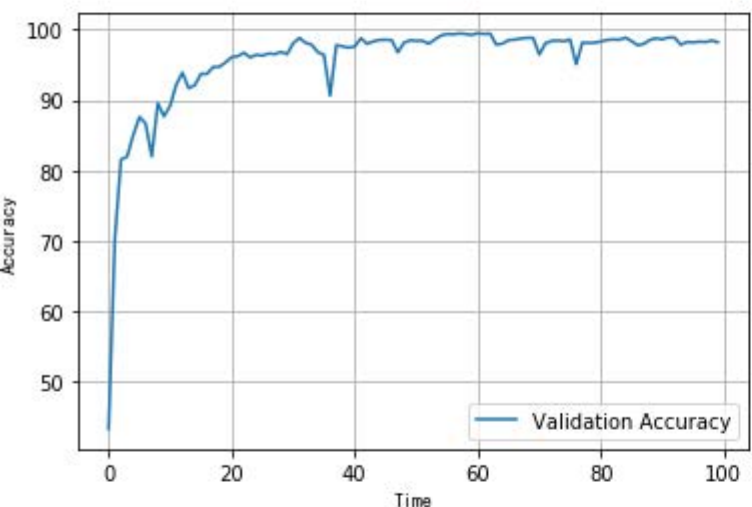
Figure 1: Model summary for GoogLeNet

Layer (type)	Output Shape	Param #
Conv2d-1	[-1, 100, 30, 30]	2,800
Conv2d-2	[-1, 150, 12, 12]	240,150
Dropout2d-3	[-1, 150, 12, 12]	0
Conv2d-4	[-1, 250, 4, 4]	337,750
Dropout2d-5	[-1, 250, 4, 4]	0
Linear-6	[-1, 200]	200,200
Linear-7	[-1, 43]	8,643
Total params: 789,543		
Trainable params: 789,543		
Non-trainable params: 0		
Input size (MB): 0.01		
Forward/backward pass size (MB): 1.08		
Params size (MB): 3.01		
Estimated Total Size (MB): 4.10		

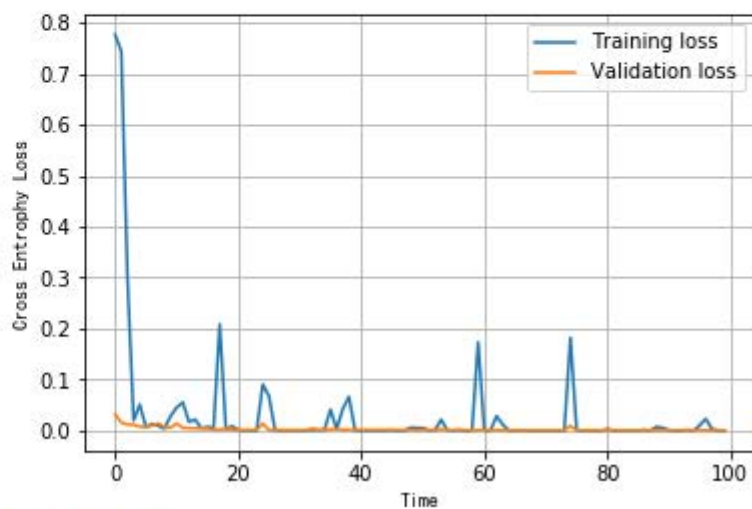
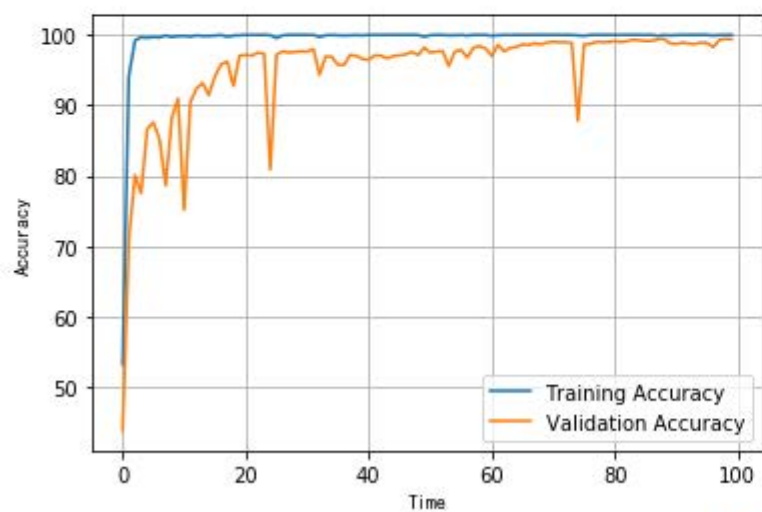
Figure 2: Model summary for A Basic CNN Net



The Basic Net



GoogLeNet-test-2



GoogLeNet-test-3