

CS-89.31: Deep Learning Generalization and Robustness

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04/27/2023

As we can see in table 2, the first neural network, having 1024 hidden units, learns about 4 times more parameters than the second one which has 256 hidden units. This is expected since the number of hidden units in the first neural network is exactly four times as many as the number of hidden units in the second. However, table 1 shows that both models have comparable performance on the dataset, with a validation error of 0.505 for the model with 1024 hidden units, compared to 0.515 for the model with 256 hidden units. In this regard, it may seem that the hidden unit trade-off might be worth it; learning fewer parameters yields almost identical error rates.

However, when we look at the computed generalization bounds, we notice that the 1024-layer neural network has much larger VC bound, L_{max} bound, and generalization bound by a factor of about 4. Therefore, the 256-layer neural network is actually more robust, suggesting that the 1024-layer neural network might have learned some spurious patterns in the data and is therefore more prone to overfitting.

Final Result	Model 1	Model 2
Training Loss	1.945	1.959
Training Margin	−0.983	−0.987
Training Error	0.468	0.484
Validation Error	0.505	0.515

TABLE 1. Results.

Metric / Hyperparameter	Model 1	Model 2
Learning Rate	0.001	
Momentum	0.9	
Batch Size	64	
Epochs	25	
Stop Condition	0.01	
Dataset	CIFAR10	
Channels	3	
Classes	10	
Hidden Units	1024	256
Frobenius ₁	19.0	10.1
Frobenius ₂	4.53	4.36
Distance ₁	4.02	4.04
Distance ₂	3.15	2.88
Spectral ₁	1.5	1.5
Spectral ₂	1.93	1.84
Fro ²	86.0	44.0
$L1^2_{max}$	1300	725
Spec Dist	15.0	13.6
Dist Spec	248	119
Spec Dist sum	263	133
Spec L1max	14.8	13.4
L1max Spec	240	115
Spec L1max sum	255	129
Dist Fro	18.2	17.6
Parameters Learned	3.16×10^6	7.89×10^5
VC bound	9.19×10^9	2.03×10^9
L1max bound	3.93×10^{10}	1.21×10^9
Computed Bound	1.88×10^{10}	4.88×10^9

TABLE 2. Result comparisons for the neural network models.