Machine Learning for Cyber-security Lab2

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I would like to introduce my work based on the sections from the notebook given us. Our work is to design a backdoor detector for BadNet trained on the YouTube Face dataset using a specific pruning defense method. The detector should take a backdoored neural network and a validation dataset as input and output a repaired BadNet which can indicate the backdoored input. I use the functions from CSAW-HackML-2020 (https://github.com/csaw-hackml/CSAW-HackML-2020) as required.

Method

We first read in 3 sets of data, valid data, test data and poisoned data. The valid data is used to pruning the network and the rest 2 data are used to test the performance of the repaired BadNets. Then we load the pre-trained BadNet based on the poisoned data “sunglasses backdoor”, this BadNet is the one that we are going to repair.

The repairing method is very simple, I calculate the average activation of each neuron in pool3 layer and sort the channels in increase order. Then I prune the conv4 layer based on the sorted index calculated before. Once a time, while testing the validation accuracy, then stop pruning when the accuracy is lower than a threshold.

Actually, I found that prune conv4 has nearly no effect, so I also try to prune conv3 layer.

Result

The accuracy valid on poisoned data and clean test data for the repaired model (prune conv4) is shown below:

	2%	4%	10%
Left channels	6	4	3
Accuracy for Clean valid data	96.28%	91.96%	87.75%
Accuracy for Clean test data	95.69%	91.04%	86.68%
Accuracy for poisoned data	99.97%	99.94%	99.92%

We can compare it to the original model:

Left channels	60
Accuracy for Clean valid data	98.65%
Accuracy for Clean test data	98.62%
Accuracy for poisoned data	100%

The accuracy valid on poisoned data and clean test data for the repaired model (prune conv3) is shown below:

	2%	4%	10%
Left channels	16	12	3
Accuracy for Clean valid data	95.75%	92.09%	84.44%
Accuracy for Clean test data	95.74%	92.13%	84.33%
Accuracy for poisoned data	100%	99.98%	77.21%

Conclusion

As we can see in the tables above, the method for prune conv4 nearly has no effect on repairing the backdoored BadNet. I think that prune conv4 using the sorting order for pool3 does not match.

The prune conv3 method works when the accuracy for clean valid data drop 10%. The accuracy for poisoned data is 77.21%, I think the method works since the accuracy for the poisoned data drops more. However, I think the effect is not enough, the mainly reason maybe that we only prune one layer. I find that many pruning methods instead of only prune one layer, they are implemented on all the layers like conv1, conv2, conv3 and conv4.

Also, I found that pruning is also not enough to solve the problem, we also need to refit the network with validate data after pruning to make a better performance. Maybe I can implement is in the future.