COMP90073 - Security Analytics

Week 11 Workshop

The purpose of this tutorial is to help you further understand adversarial training as well as its limitation.

Instructions:

1. Run "mnist_tutorial_tf.py" in the subfolder of "cleverhans_tutorials", and test whether the adversarially trained mode (model2) is also robust against adversarial samples generated by the indiscriminate C&W L2 attack.

Hint: check how indiscriminate C&W L2 attack is implemented in "mnist tutorial cw.py".

Expected result:

- 1. The model trained on clean examples (model1) is not robust against adversarial samples generated by the Fast Gradient Sign Method (FGSM) the accuracy on adversarial samples is around 10%;
- 2. The model trained on adversarial samples (model2) generated by FGSM is much more robust the accuracy on adversarial samples increases to over 95%;
- 3. However, since "model2" is trained on adversarial samples generated by FGSM (a relatively weak form of adversarial attack), it is not robust against adversarial samples generated by the indiscriminate C&W L2 attack the accuracy goes back to around 10%.

ASSIGNMENT Project Exam Help (The percentages may differ on your machine)

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Test accuracy on adversarial examples: 0.1000
Repeating the process, using adversarial training
Test accuracy on legitimate star ples: 0.9520 tutores. Com

[INFO 2019-09-07 02:33:14,742 cleverhans] Constructing new graph for attack CarliniWagnerL2
                                                      ('Running CWL2 attack on instance %s of %s', 0, 1000)
[DEBUG 2019-09-07 02:33:15,849 cleverhans]
[DEBUG 2019-09-07 02:33:16,123 cleverhans]

[DEBUG 2019-09-07 02:33:17,052 cleverhans]

[DEBUG 2019-09-07 02:33:20,783 cleverhan]

[DEBUG 2019-09-07 02:33:24,493 cleverhan]
                                                         Binary search step 0 of 1
                                                           Iteration 0 of 50: loss=3.99e+04 l2=0 f=-0.389
                                                           tteration 5 of 56: 105:-2.8e+04 12=4.16 f=-0.367
tteration 1: 0 of 50: 105:2.8e+04 12=7.42 f=-0.362
Iteration 15 of 50: 1055=2e+04 12=8.8 f=-0.363
DEBUG 2019-09-07 02:33:28,583 cleverhans
DEBUG 2019-09-07 02:33:32,625 cleverhans
                                                            Iteration 20 of 50: loss=1.74e+04 l2=8.9 f=-0.371
DEBUG 2019-09-07 02:33:36,749 cleverhans
                                                            Iteration 25 of 50: loss=1.58e+04 l2=8.88 f=-0.376
DEBUG 2019-09-07 02:33:41,072 cleverhans
                                                            Iteration 30 of 50: loss=1.48e+04 l2=8.65 f=-0.378
DEBUG 2019-09-07 02:33:46,142 cleverhans
                                                            Iteration 35 of 50: loss=1.35e+04 l2=8.51 f=-0.376
[DEBUG 2019-09-07 02:33:50,073 cleverhans
[DEBUG 2019-09-07 02:33:53,893 cleverhans
                                                            Iteration 40 of 50: loss=1.22e+04 l2=8.63 f=-0.377
                                                           Iteration 45 of 50: loss=1.12e+04 l2=8.41 f=-0.378
                                                         Successfully generated adversarial examples on 881 of 1000 instances.
DEBUG 2019-09-07 02:33:57,266 cleverhans
[DEBUG 2019-09-07 02:33:57,267 cleverhans]
                                                          Mean successful distortion: 2.667
Test accuracy on adv. examples generated by C&W: 0.1250
 ress any key to continue . . .
```