

Team Teaflow Non-targeted / Targeted Attack Summary

Caspar Zhang
casparz@gmail.com

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In CAAD 2018 competition, our submissions on Non-targeted & Targeted Attack task are both based on Basic Iterative Method(BIM).

1. Gradient is smoothed spatially using Gaussian filter:

$$\begin{aligned} X_0^{adv} &= X \\ X_{N+1}^{adv} &= \text{Clip}_{X,\epsilon}(X_N^{adv} + Y_N) \\ Y_N &= \alpha \text{sign}(\nabla_X J(X_N^{adv}, Y_{true})) \end{aligned}$$



$$\begin{aligned} X_0^{adv} &= X \\ X_{N+1}^{adv} &= \text{Clip}_{X,\epsilon}(X_N^{adv} + Y_N) \\ Y_N &= \alpha \text{sign}(h_{smooth} * \nabla_X J(X_N^{adv}, Y_{true})) \end{aligned}$$

The parameters of Gaussian filter are set separately for the two tasks:

- 1) non-targeted attack
 $size(kernel) = 7$
 $\sigma = 2$
- 2) targeted attack
 $size(kernel) = 6$
 $\sigma = 4$

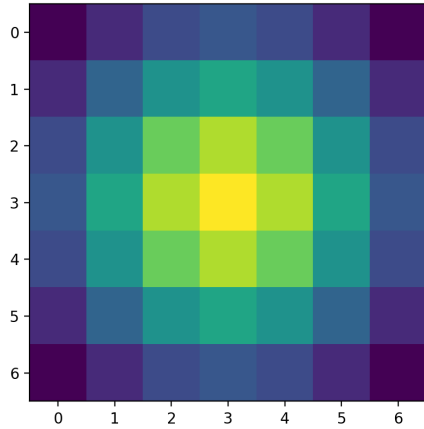


Figure 1 non-targeted attack

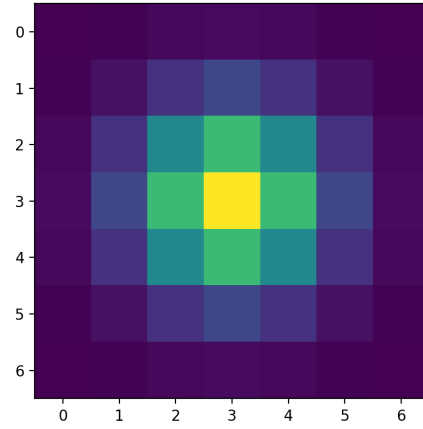


Figure 1 targeted attack

2. After some comparison tests, we decided to use the following models for ensembling:

1) non-targeted attack

- i. adv_inception_v3_2017_08_18
- ii. ens4_adv_inception_v3_2017_08_18
- iii. adv_inception_resnet_v2_2017_12_18

2) targeted attack

- i. ens_adv_inception_resnet_v2_2017_08_18
- ii. adv_inception_v3_2017_08_18
- iii. inception_v3_2016_08_28

3. Engineering direction improvement:

- 1) Using TensorRT to speedup TensorFlow inference, try to perform more iterations under the time constraint
- 2) Compiling Tensorflow with SSE, AVX & AVX2 instructions support

4. Benchmark

For benchmarking, we use all images in the DEV dataset. For evaluation, we used the top-2 NIPS 2017 defense solutions and also some baselines (fgsm & adv_inception_resnet).

	TsAIL	iyswim	adv_inception_resnet
fgsm	16.8%	39.0%	39.5%
teafLOW	93.7%	64.3%	71.1%

Table 1 Non-targeted Attack Success Rate

	TsAIL	iyswim	adv_inception_resnet
BIM	1.2%	2.4%	2.8%
teaflow	46.7%	16.4%	19.6%

Table 2 Targeted Attack Success Rate