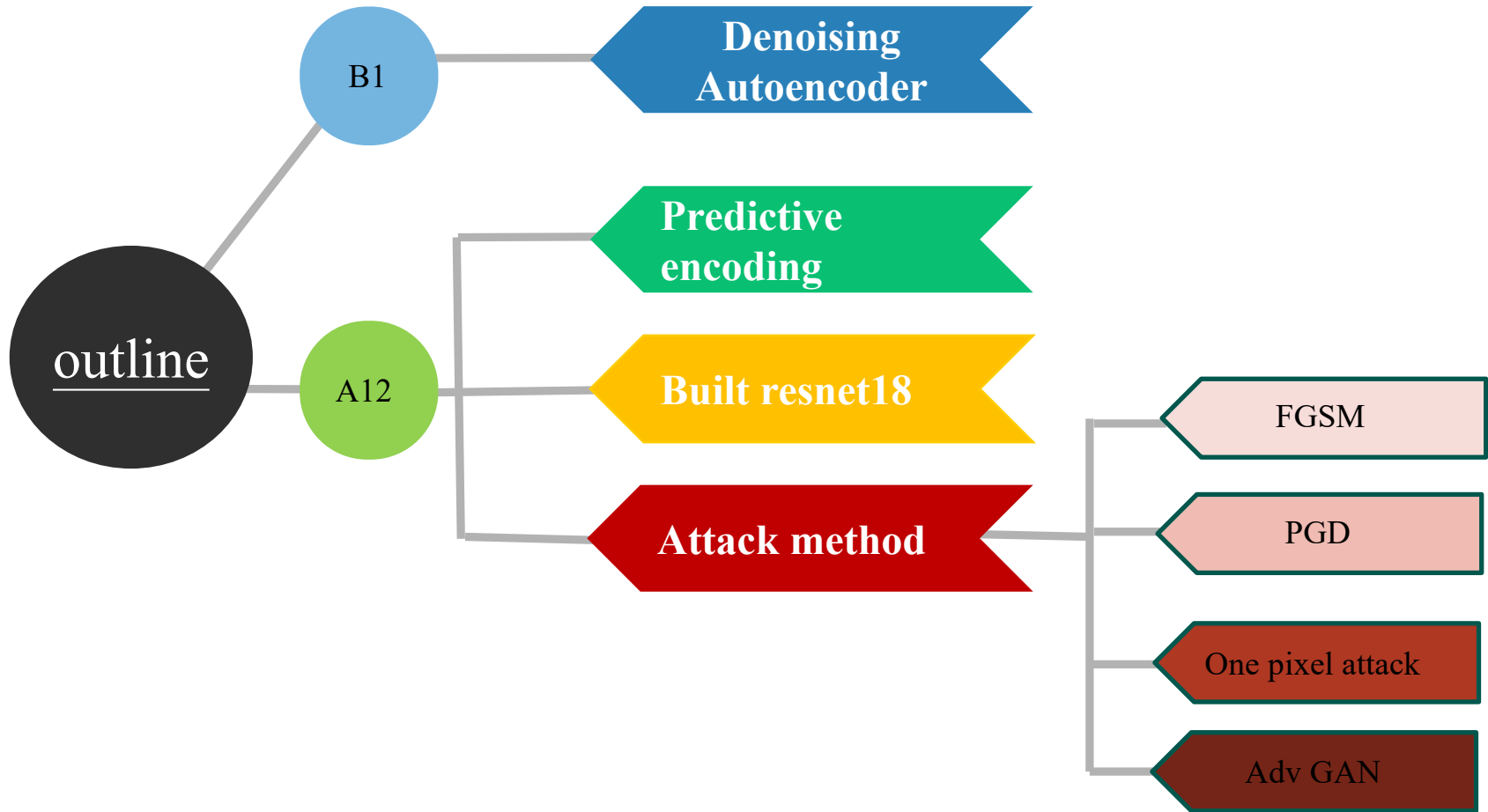


Result report

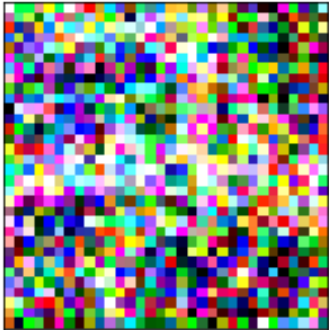
毛柏元 Mao po yuan
E-mail:zxc596666123@gmail.com



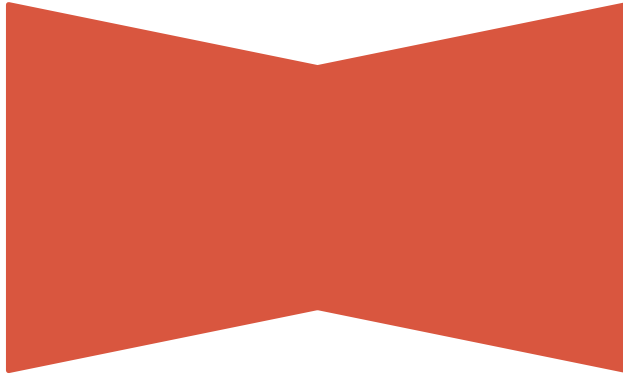
B1 Reconstruct images from CIFAR 10

Denoising Autoencoder

Reference <https://codahead.com/blog/a-denoising-autoencoder-for-cifar-datasets>

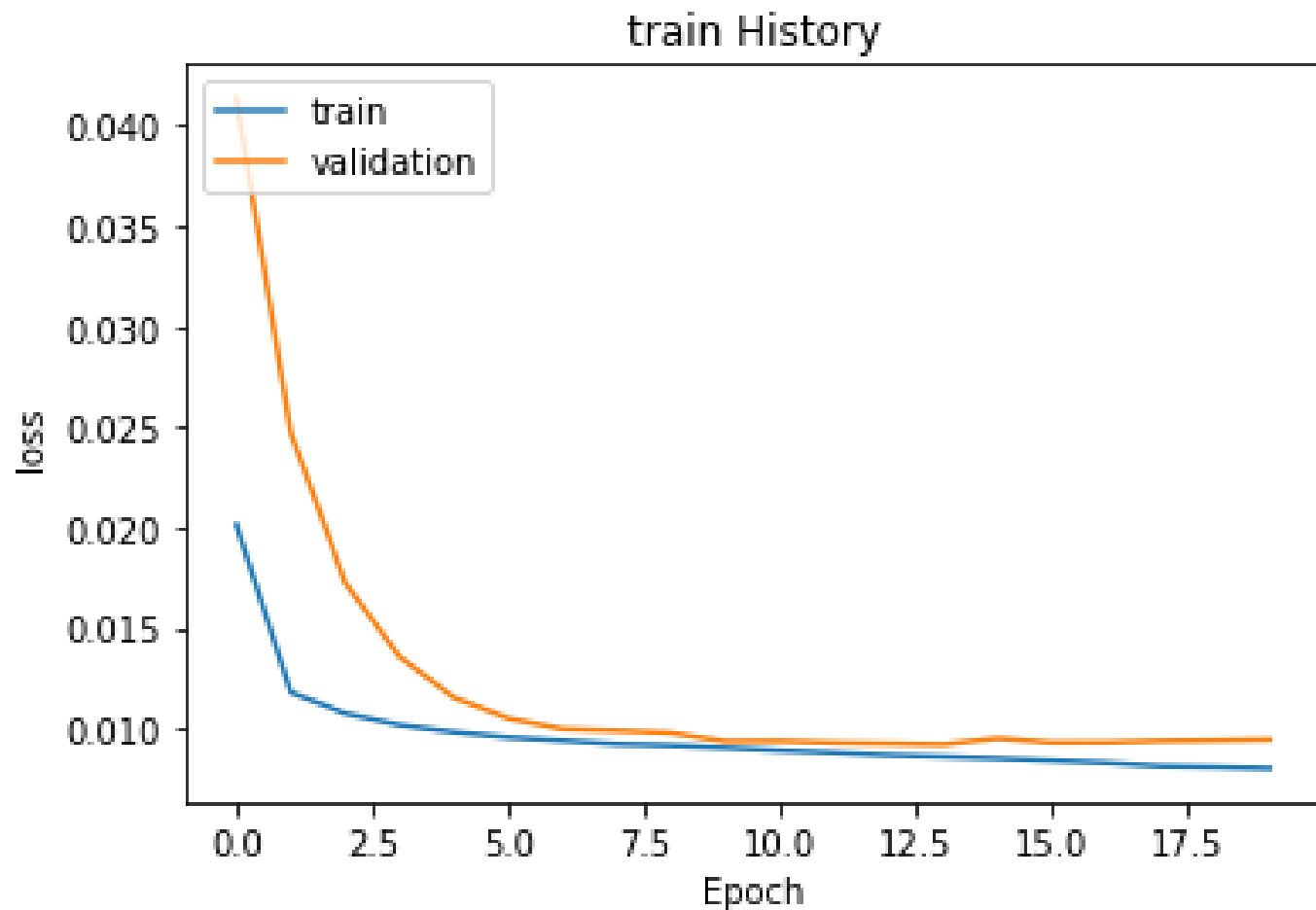


Noise picture
(clean picture with random noise)

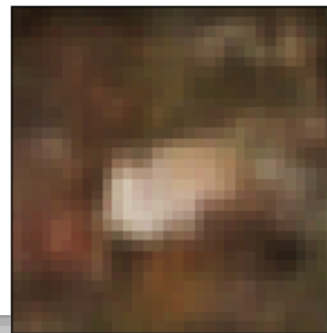
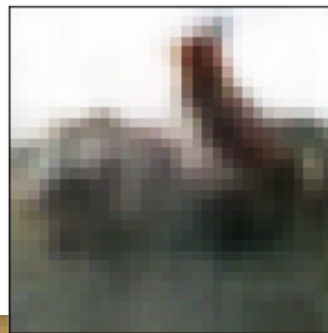
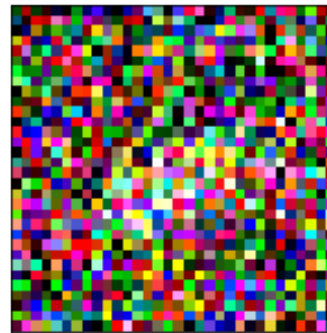
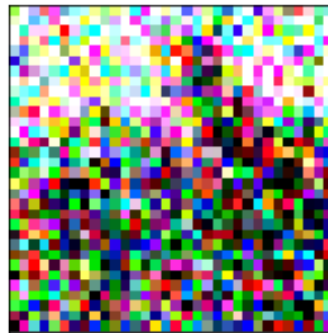
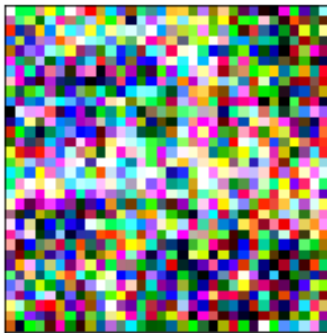
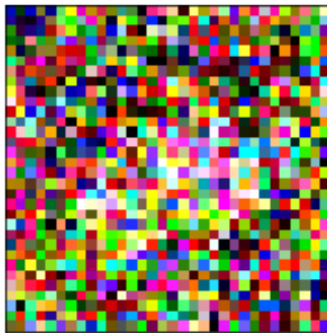
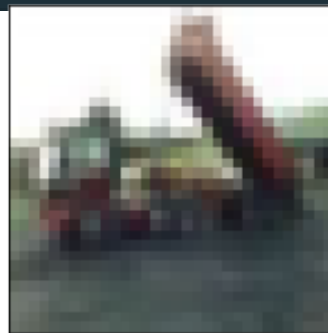


decode picture

history

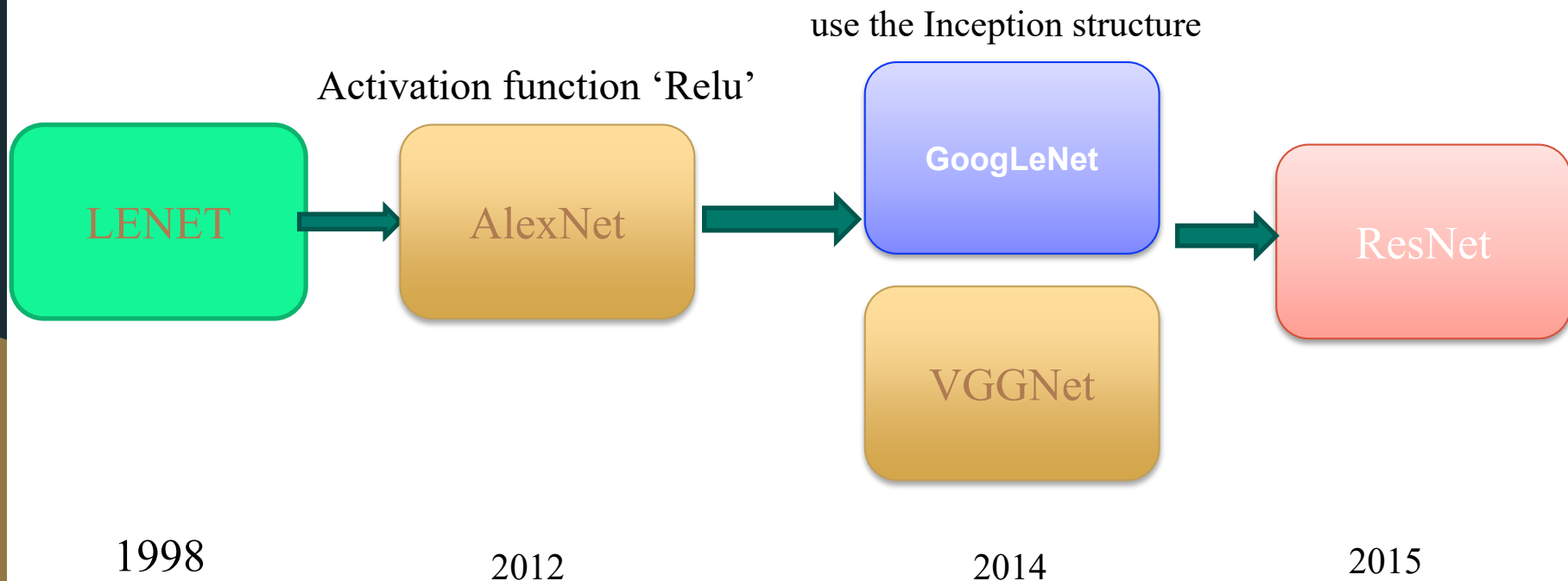


result



All create a predictive encoding
Model on CIFAR10 and
attack it to verify its robustness

predictive encoding



Reference : [\(AlexNet · VGG · Inception · ResNet\)+Keras Coding](#)



Fast Gradient Sign Method (FGSM)

white box attack

Input : image and Model
Output : Perturbation



FGSM

$$x^* \leftarrow x^0 - \varepsilon \Delta x$$

$$\Delta x = \begin{bmatrix} \text{sign}(\partial L / \partial x_1) \\ \text{sign}(\partial L / \partial x_2) \\ \vdots \\ \vdots \end{bmatrix}$$

x^* : picture with perturbation

L : loss function

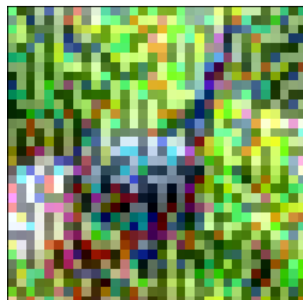
ε : limit of perturbation

x^0 : original picture

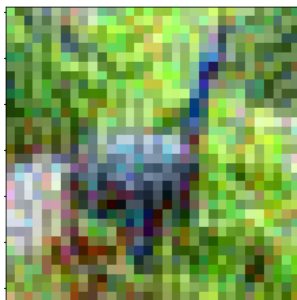
ϵ	0.2	0.1	0.05	0.01
Success rate	66.03%	68.44%	69.37%	65.83%
Success Robustness	0.404	0.203	0.1019	0.02



Bird 99.9%



4.9e-5% bird
99.9% frog



2.0e-2% bird
65.5% frog



5.1e-2% bird
42.5% cat



12.3% bird
29.3% cat

PGD ATTACK

white box attack

Input : image and Model
Output : Perturbation

Reference : Towards Deep Learning Models Resistant to Adversarial Attacks
<https://arxiv.org/pdf/1706.06083.pdf>

PGD

$$x^n \leftarrow x^{n-1} - a\Delta x^{n-1}$$

$$\Delta x^{n-1} = \begin{bmatrix} \text{sign}(\partial L / \partial x_1) \\ \text{sign}(\partial L / \partial x_2) \\ \vdots \\ \vdots \end{bmatrix}$$

$$\varepsilon \geq \sum a \Delta x^{n-1}$$

x^n : picture with all perturbation
at epochs n

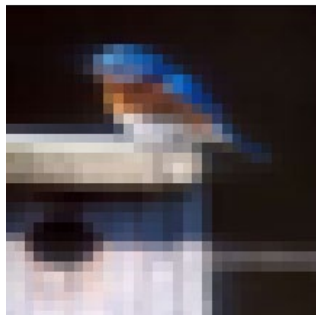
a : each epochs step perturbation

ε : limit of perturbation

ε	0.1	0.1
Epochs	10	20
α (each epochs step perturbation)	0.01	0.001
Success Robustness	0.0314	0.0149
Success rate	70.19	70.2%

Result

Original image



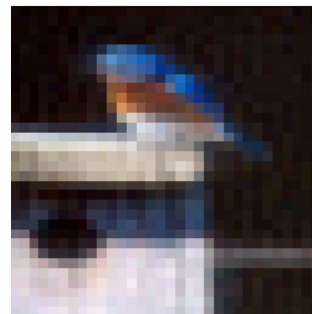
Original 99.9% bird



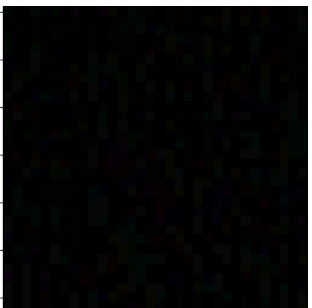
perturbation



100% truck
3.2e-14% bird



Original 99.9% truck



99.9% bird
1.14e-10% truck

ONE pixel attack

black box attack

Input : image
Output : Perturbation

Reference :One pixel attack

<https://arxiv.org/abs/1710.08864>

Difference from others attack

$$\underset{e(\mathbf{x})^*}{\text{maximize}} \quad f_{adv}(\mathbf{x} + e(\mathbf{x}))$$

$$\text{subject to} \quad \|e(\mathbf{x})\| \leq L$$

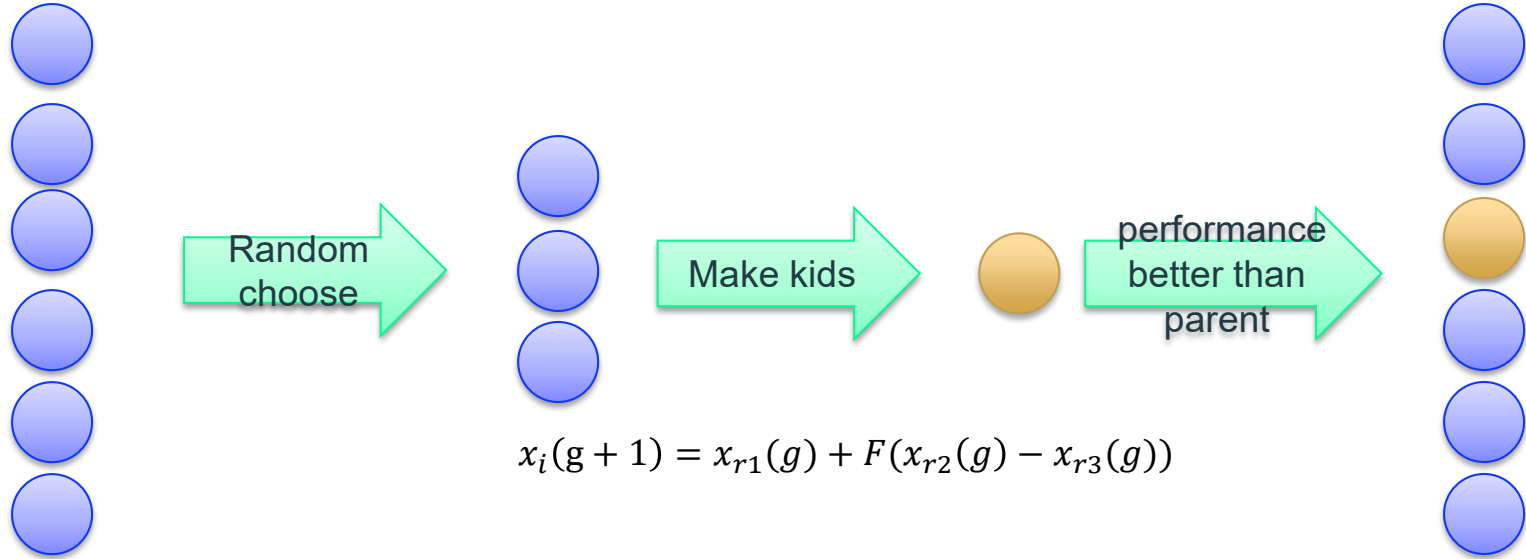
Other attacks

$$\underset{e(\mathbf{x})^*}{\text{maximize}} \quad f_{adv}(\mathbf{x} + e(\mathbf{x}))$$

$$\text{subject to} \quad \|e(\mathbf{x})\|_0 \leq d,$$

One pixel attack

Method: Differential Evolution



Parents(x , y , R,G,B)

New Parents(x , y , R,G,B)

Result



Original 81.8% automobile
After 34% automobile
37.8% truck



Original 78.8% deer
After 6.7% deer
91.6% cat



Original 81.9% cat
After 8.5% cat
90.6% frog

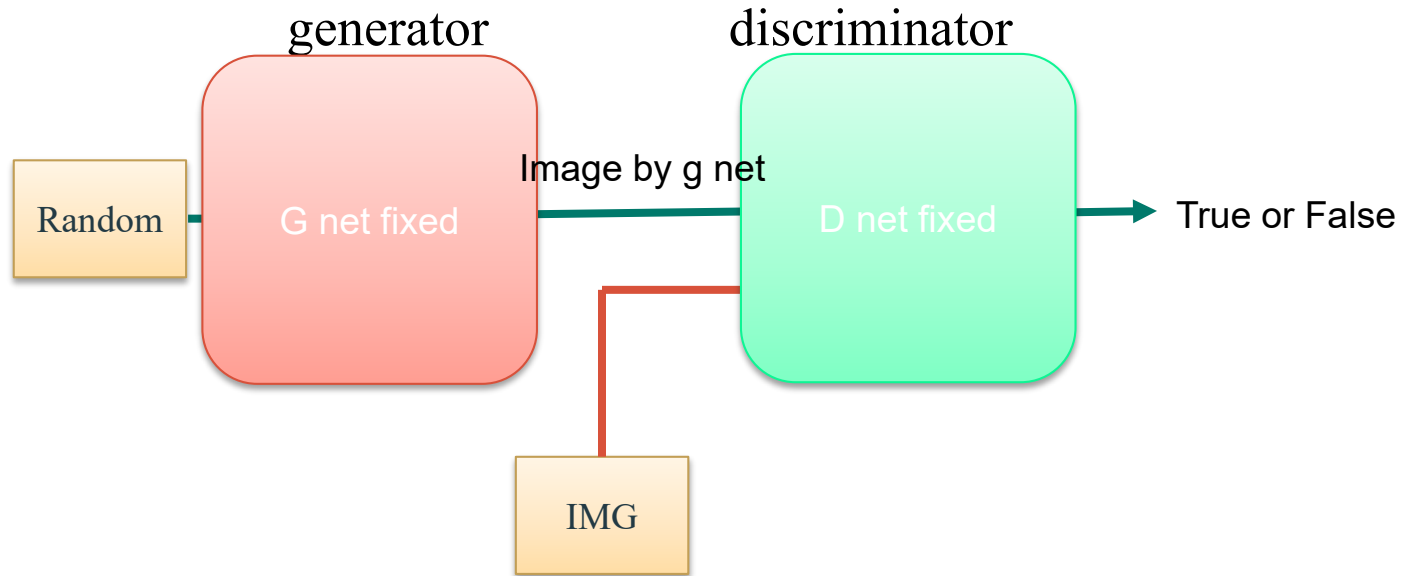
ADV GAN

white box attack

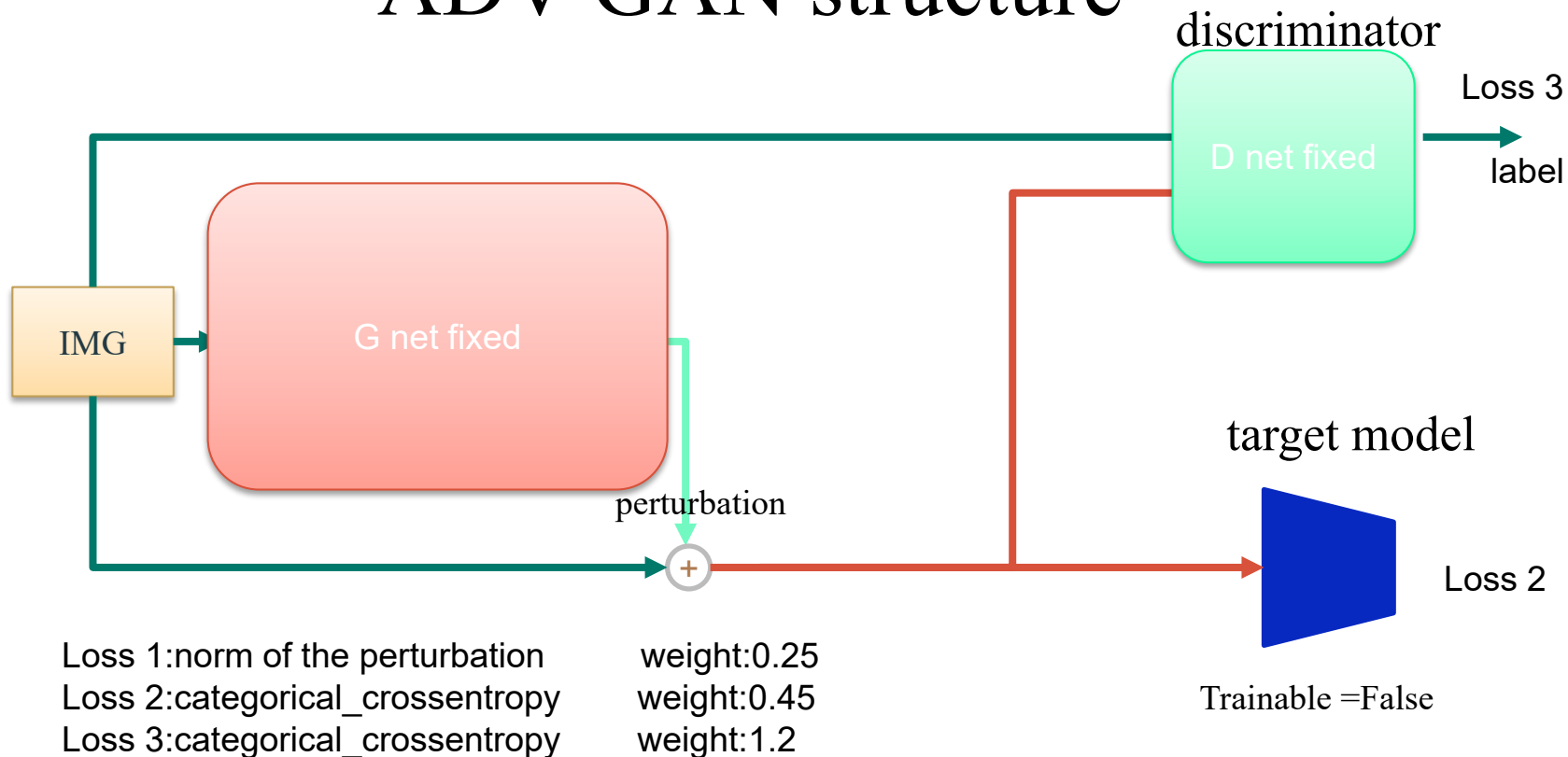
Input : image and Model
Output : Perturbation

Reference :ADVGAN <https://arxiv.org/abs/1801.02610>

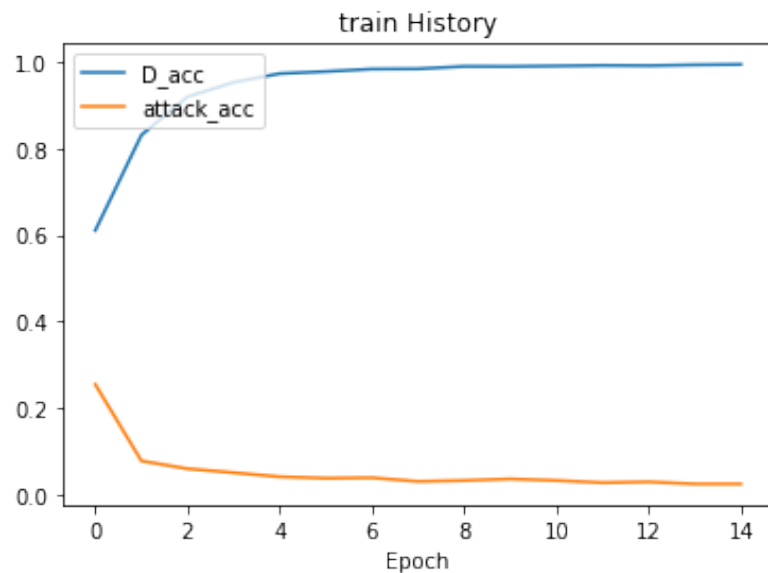
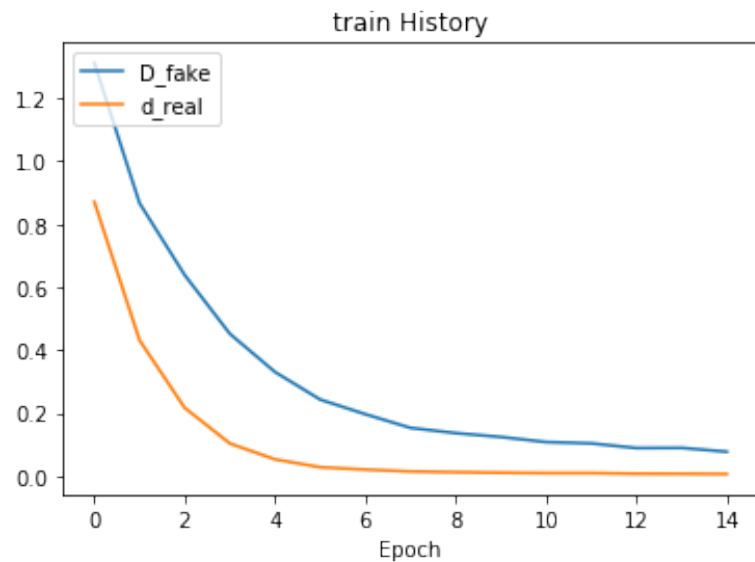
GAN



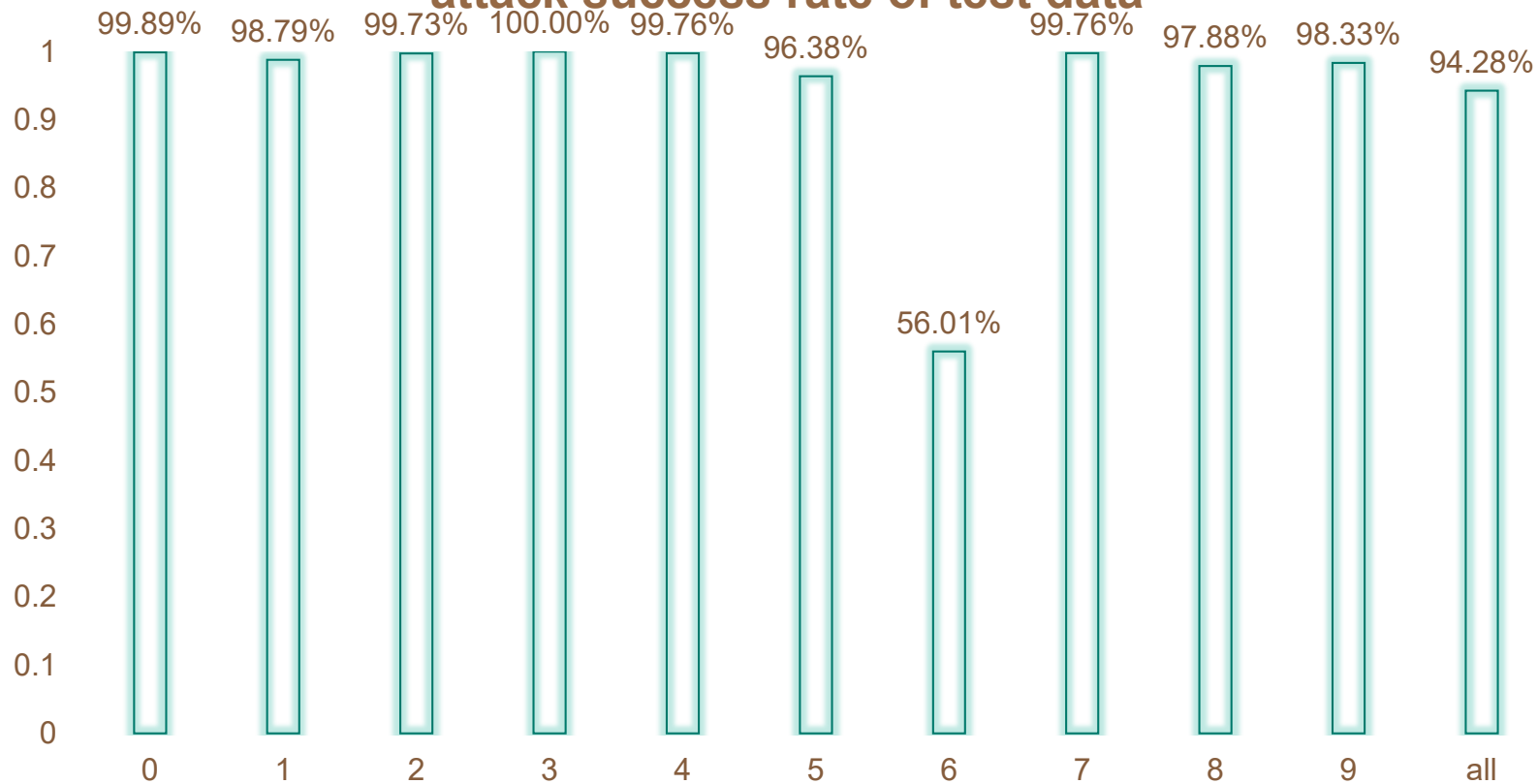
ADV GAN structure



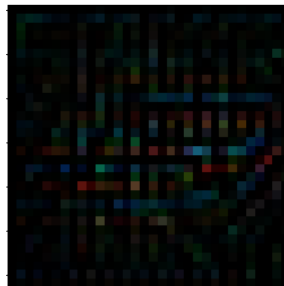
history



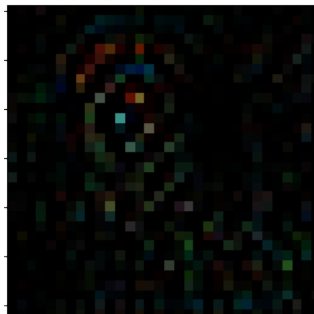
attack success rate of test data



Result



100% ship
2.68e-11% frog



99.99%dog
6.47e-7% truck



Comparison

Attack method	FGSM	PDG	one pixel attack	Adv gan
Black box or White box	White box	White box	Black box	White box
Success rate	69.37%	70.2%	53%	96.24%
Robustness	0.1019	0.0149		0.0967
advantage	Fast	1. Stable success rate 2. Get the less perturbation to mis- lead the model	Do not need the model's detail	1.high success rate 2.Easy to use



Thank you for listening