

МИНОБРНАУКИ РОССИИ

Федеральное государственное бюджетное образовательное учреждение высшего образования

«МИРЭА – Российский технологический университет» РТУ МИРЭА

ИКБ направление «Киберразведка и противодействие угрозам с применением технологий искусственного интеллекта» 10.04.01

Кафедра КБ-4 «Интеллектуальные системы информационной безопасности»

Лабораторная работа №1

по дисциплине

«Анализ защищенности систем искусственного интеллекта»

Группа: ББМО-01-22

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Проверил: Спирин А.А. • Клонируем репозиторий.

```
In [1]: !git clone https://github.com/ewatson2/EEL6812_DeepFool_Project.git

Cloning into 'EEL6812_DeepFool_Project'...
    remote: Enumerating objects: 96, done.
    remote: Counting objects: 100% (3/3), done.
    remote: Compressing objects: 100% (2/2), done.
    remote: Total 96 (delta 2), reused 1 (delta 1), pack-reused 93
    Receiving objects: 100% (96/96), 33.99 MiB | 15.11 MiB/s, done.
    Resolving deltas: 100% (27/27), done.
```

• Сменим директорию

```
%cd /content/EEL6812_DeepFool_Project
```

/content/EEL6812_DeepFool_Project

• Импортируем библиотеки.

```
import numpy as np
import os
import json, torch
from torch.utils.data import DataLoader, random_split
from torchvision import datasets, models
from torchvision.transforms import transforms
from models.project_models import FC_500_150, LeNet_CIFAR, LeNet_MNIST, Net
from utils.project_utils import get_clip_bounds, evaluate_attack, display_attack
```

• Установка случайного значения, равного номеру в списке (39).

```
In [4]:
    rand_seed = 39
    np.random.seed(rand_seed)
    torch.manual_seed(rand_seed)

    use_cuda = torch.cuda.is_available()
    device = torch.device('cuda' if use_cuda else 'cpu')
```

• Загрузка датасета MNIST.

```
mnist_mean = 0.5
mnist_std = 0.5
mnist_std = 0.5
mnist_dim = 28
mnist_min, mnist_max = get_clip_bounds(mnist_mean, mnist_std, mnist_dim)
mnist_min = mnist_min.to(device)
mnist_min = mnist_min.to(device)
mnist_tf = transforms.Compose([ transforms.ToTensor(), transforms.Normalize( mean=mnist_mean, std=mnist_std)])
mnist_tf_train = transforms.Compose([ transforms.RandomHorizontalFlip(), transforms.ToTensor(), transforms.Normalize( mean=mnist_mean, std=mnist_std)])
mnist_tf_inv = transforms.Compose([ transforms.Normalize( mean=0.0, std=np.divide(1.0, mnist_std)), transforms.Normalize( mean=np.multiply(-1.0, mnist_std)), std=1.0])
mnist_temp = datasets.MNIST(root='datasets/mnist', train=True, download=True, transform=mnist_tf_train)
mnist_train, mnist_val = random_split(mnist_temp, [50000, 10000])
mnist_test = datasets.MNIST(root='datasets/mnist', train=False, download=True, transform=mnist_tf)
```

• Загрузка датасета CIFAR-10.

```
cifar_mean = [0.491, 0.482, 0.447]
cifar_std = [0.202, 0.199, 0.201]
cifar_std = [0.202, 0.199, 0.201]
cifar_min = cifar_min. cifar_max = get_clip_bounds(cifar_mean, cifar_std, cifar_dim)
cifar_min = cifar_min.to(device)
cifar_max = cifar_max.to(device)
cifar_max = cifar_max.to(device)
cifar_tf = transforms.Compose([ transforms.Normalize( mean-cifar_mean, std-cifar_std)])
cifar_tf_train = transforms.Compose([ transforms.RandomCrop( size=cifar_dim, padding=4), transforms.RandomHorizontalFlip(), transforms.ToTensor(), transforms.Normalize( mean-cifar_mean, std-cifar_std)])
cifar_tf_inv = transforms.Compose([ transforms.Normalize( mean-[0.0, 0.0, 0.0], std-np.divide(1.0, cifar_std)), transforms.Normalize( mean-np.multiply(-1.0, cifar_mean), std-[1.0, 1.0])])
cifar_temp = datasets.CIFAR10(root='datasets/cifar-10', train=True, download=True, transform=cifar_tf_train)
cifar_train, cifar_val = random_split(cifar_temp, [40000, 10000])
cifar_test = datasets.CIFAR10(root='datasets/cifar-10', train=False, download=True, transform=cifar_tf)
cifar_classes = ['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'truck']
```

• Выполнение настройки и загрузки DataLoader.

```
batch_size = 64
workers = 4
mnist_loader_train = DataLoader(mnist_train, batch_size=batch_size, shuffle=True, num_workers=workers)
mnist_loader_val = DataLoader(mnist_val, batch_size=batch_size, shuffle=False, num_workers=workers)
mnist_loader_test = DataLoader(mnist_test, batch_size=batch_size, shuffle=False, num_workers=workers)
cifar_loader_train = DataLoader(cifar_train, batch_size=batch_size, shuffle=False, num_workers=workers)
cifar_loader_val = DataLoader(cifar_val, batch_size=batch_size, shuffle=False, num_workers=workers)
cifar_loader_test = DataLoader(cifar_test, batch_size=batch_size, shuffle=False, num_workers=workers)
```

• Настройка обучающей модель.

```
train_model = True

epochs = 50
epochs_nin = 100

lr = 0.004
lr_nin = 0.01
lr_scale = 0.5

momentum = 0.9

print_step = 5

deep_batch_size = 10
deep_num_classes = 10
deep_overshoot = 0.02
deep_max_iters = 50

deep_args = [deep_batch_size, deep_num_classes, deep_overshoot, deep_max_iters]
if not os.path.isdir('weights/deepfool'): os.makedirs('weights/fgsm', exist_ok=True)

if not os.path.isdir('weights/fgsm'): os.makedirs('weights/fgsm', exist_ok=True)
```

• Загрузка и оценка стойкости модели Network-In-Network Model к FGSM и DeepFool атакам на основе датасета CIFAR-10.

```
fgsm_eps = 0.2
model = Net().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_nin.pth', map_location=torch.device('cpu')))
evaluate_attack('cifar_nin_fgsm.csv', 'results', device, model, cifar_loader_test, cifar_min, cifar_max, fgsm_eps, is_fgsm=True)
print('')
evaluate_attack('cifar_nin_deepfool.csv', 'results', device, model, cifar_loader_test, cifar_min, cifar_max, deep_args, is_fgsm=False)
if device.type == 'cuda': torch.cuda.empty_cache()

FGSM Test Error : 81.29%
FGSM Robustness : 1.77e-01
FGSM Time (All Images) : 0.67 s
FGSM Time (Per Image) : 67.07 us

DeepFool Test Error : 93.76%
DeepFool Time (All Images) : 185.12 s
DeepFool Time (All Images) : 185.12 s
DeepFool Time (Per Image) : 18.51 ms
```

• Загрузим и оценка стойкость модели LeNet к FGSM и DeepFool атакам на основе датасета CIFAR-10.

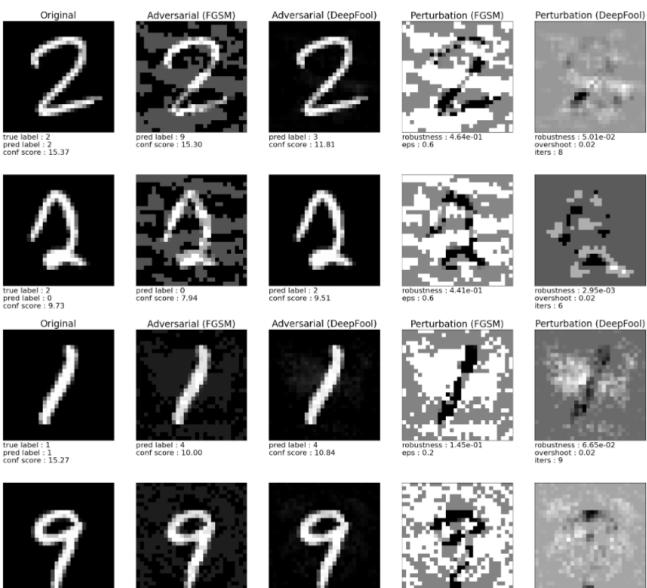
```
fgsm_eps = 0.1
model = LeNet_CIFAR().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_lenet.pth', map_location=torch.device('cpu')))
evaluate_attack('cifar_lenet_fgsm.csv', 'results', device, model, cifar_loader_test, cifar_min, cifar_max, fgsm_eps, is_fgsm=True)
print('')
evaluate_attack('cifar_lenet_deepfool.csv', 'results', device, model, cifar_loader_test, cifar_min, cifar_max, deep_args, is_fgsm=False)
if device.type == 'cuda': torch.cuda.empty_cache()

FGSM Test Error : 91.71%
FGSM Robustness : 8.90e-02
FGSM Time (All Images) : 0.40 s
FGSM Time (Per Image) : 40.08 us

DeepFool Test Error : 87.81%
DeepFool Robustness : 1.78e-02
DeepFool Time (All Images) : 73.27 s
DeepFool Time (Per Image) : 7.33 ms
```

• Выполним оценку атакующих примеров для сетей.

```
#LeNet
fgsm_eps = 0.6
model = LeNet_MNIST().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_lenet.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_
if device.type == 'cuda': torch.cuda.empty_cache()
#FCNet
fgsm eps = 0.2
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_
if device.type == 'cuda': torch.cuda.empty_cache()
#Network-in-Network
fgsm eps = 0.2
model = Net().to(device)
model.load state dict(torch.load('weights/clean/cifar nin.pth'))
display attack(device, model, cifar test, cifar tf inv, cifar min, cifar max, fgsm eps, deep args, has
if device.type == 'cuda': torch.cuda.empty_cache()
#LeNet CIFAR-10
fgsm_eps = 0.1
model = LeNet_CIFAR().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_lenet.pth'))
display attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_
if device.type == 'cuda': torch.cuda.empty cache()
```





true label : 9 pred label : 9 conf score : 18.12



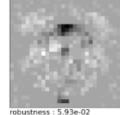
pred label : 4 conf score : 14.10



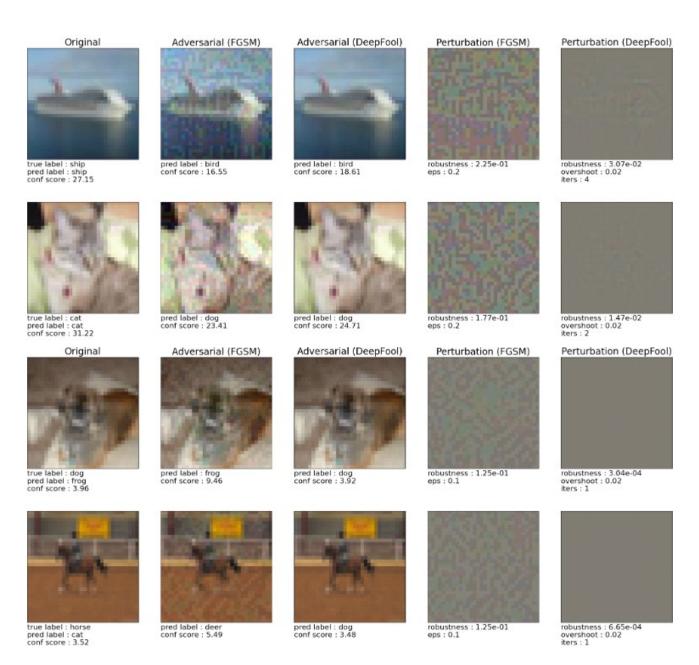
pred label : 4 conf score : 13.07



robustness: 1.61e-01 eps: 0.2

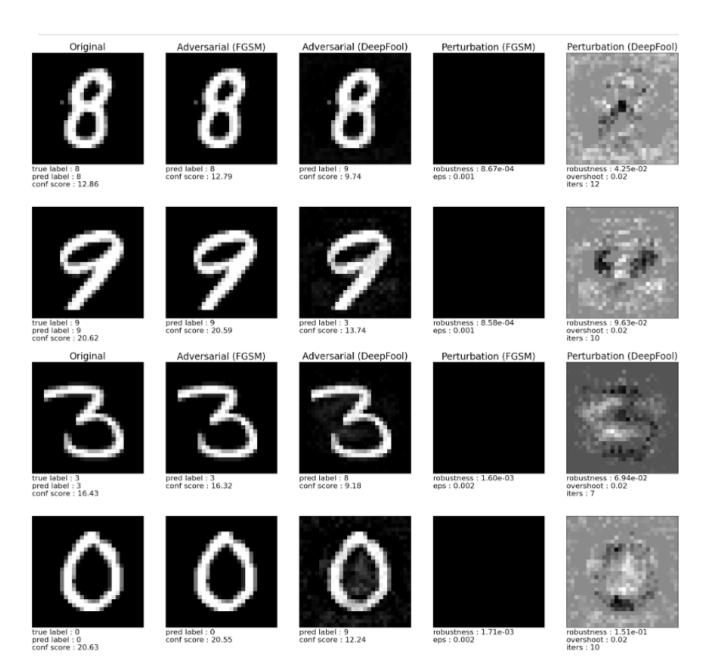


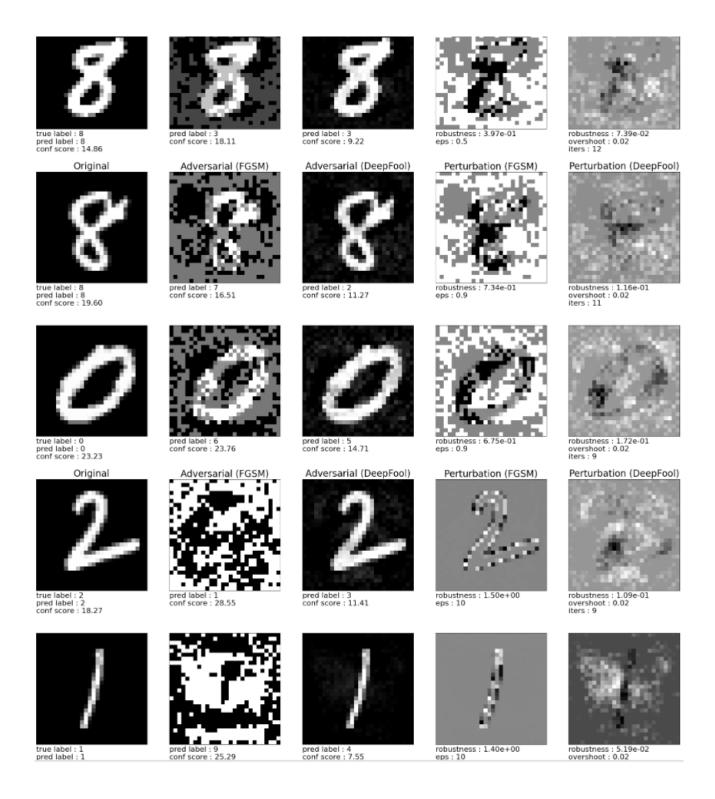
robustness : 5.93e-02 overshoot : 0.02 iters : 9



Отражаем отличия для fgsm_eps=(0.001, 0.02, 0.5, 0.9, 10).

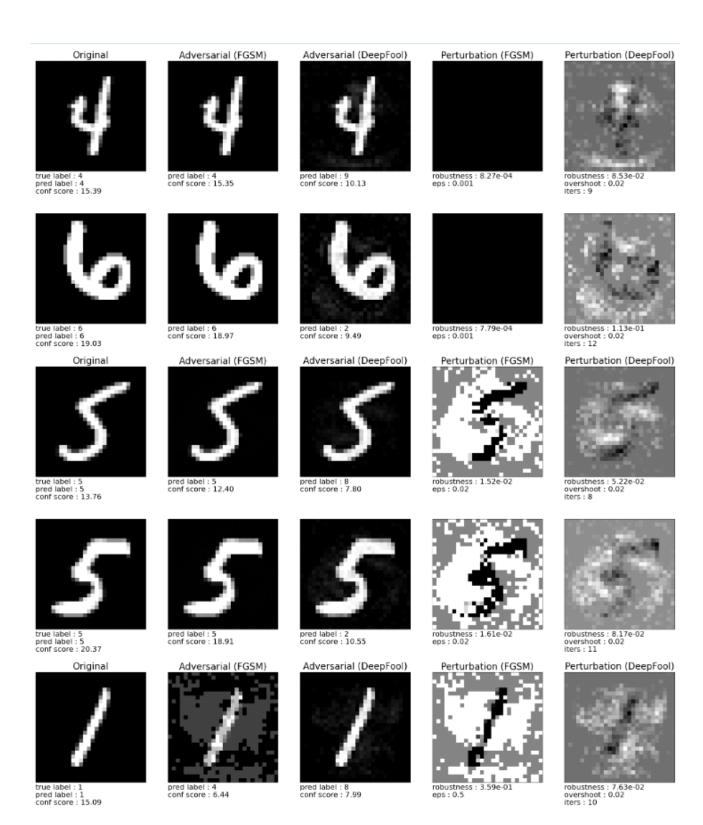
```
fgsm eps = 0.001
model = FC 500 150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.002
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.5
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_i
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.9
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_i
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 10
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_
if device.type == 'cuda': torch.cuda.empty_cache()
```

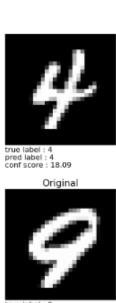




• Проверим влияние параметра fgsm_eps для FC на датасете MNIST.

```
fgsm_eps = 0.001
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, :
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.02
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, :
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.5
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, :
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.9
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_fi_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, :
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 10
model = FC_500_150().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_fc.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, :
if device.type == 'cuda': torch.cuda.empty_cache()
```

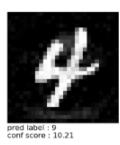


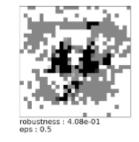


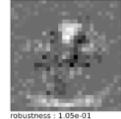


pred label: 7 conf score: 20.71

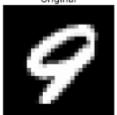








robustness: 1.05e-01 overshoot: 0.02 iters: 9



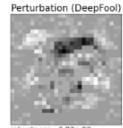
true label : 9 pred label : 9 conf score : 13.50



pred label : 4 conf score : 11.36



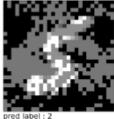
robustness eps: 0.9



robustness : 3.83e-02 overshoot : 0.02 iters : 10



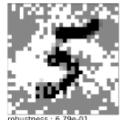
true label : 5 pred label : 5 conf score : 16.44



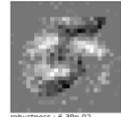
pred label : 2 conf score : 14.07



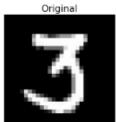
pred label : 8 conf score : 8.73



robustness: 6.79e-01 eps: 0.9



robustness : 6.39e-02 overshoot : 0.02 iters : 8



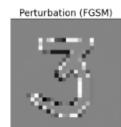
true label : 3 pred label : 3 conf score : 22.63



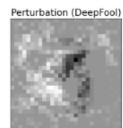
pred label : 5 conf score : 61.76



pred label : 5 conf score : 14.52



robustness: 1.43e+00 eps: 10



robustness : 9.35e-02 overshoot : 0.02 iters : 10



true label : 0 pred label : 0 conf score : 11.48



pred label : 7 conf score : 56.61



pred label : 7 conf score : 8.74



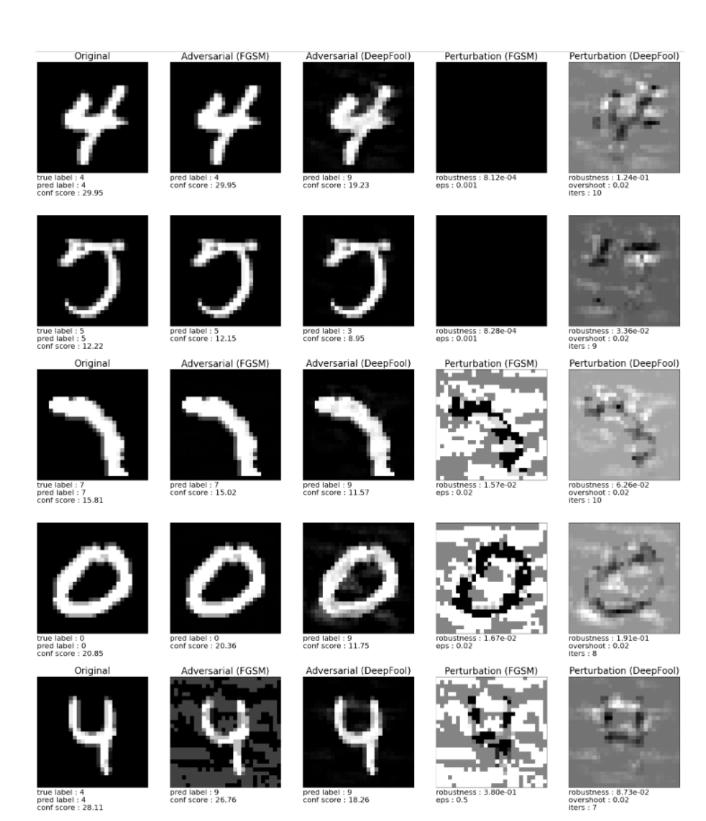
robustness : 1.56e+00 eps : 10

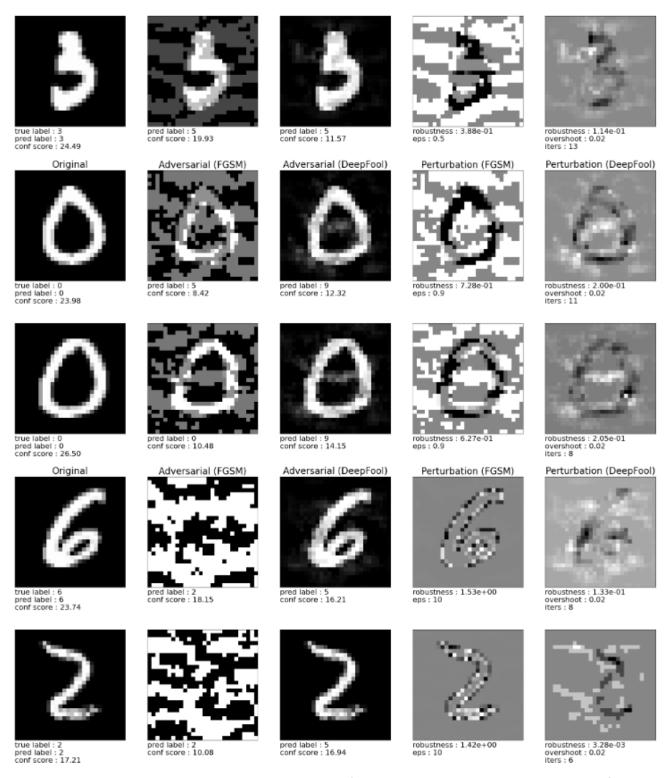


robustness: 4.94e-02 overshoot: 0.02 iters: 10

• Проверим влияние параметра fgsm_esp для LeNet на датасете MNIST.

```
fgsm_eps = 0.001
model = LeNet_MNIST().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_lenet.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_no
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.02
model = LeNet_MNIST().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_lenet.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnist_te
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.5
model = LeNet_MNIST().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_lenet.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnis
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.9
model = LeNet_MNIST().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_lenet.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nor
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 10
model = LeNet_MNIST().to(device)
model.load_state_dict(torch.load('weights/clean/mnist_lenet.pth'))
display_attack(device, model, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnist_test, mnist_tf_inv, mnist_min, mnist_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model, mnist_test, mnis
if device.type == 'cuda': torch.cuda.empty_cache()
```





• Проверим влияние параметра fgsm_esp для NiN на датасете Cifar-

10.

```
fgsm_eps = 0.001
model = Net().to(device)
 model.load_state_dict(torch.load('weights/clean/cifar_nin.pth'))
display attack(device, model, cifar test, cifar tf inv, cifar min, cifar max, fgsm eps, deep args, has labels=False, 12 nor
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.02
model = Net().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_nin.pth'))
display_attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_labels=False, 12_nometric files and files are files are files and files are files a
if device.type == 'cuda': torch.cuda.empty cache()
model = Net().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_nin.pth'))
display_attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_labels=False, 12_nor
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.9
model = Net().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_nin.pth'))
display\_attack (device, model, cifar\_test, cifar\_tf\_inv, cifar\_min, cifar\_max, fgsm\_eps, deep\_args, has\_labels = \textit{False}, 12\_noillowed by the compact of 
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 10
model = Net().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_nin.pth'))
display_attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_labels=False, 12_normax, fgsm_eps, deep_args, has_labels=False, fgsm_eps, deep_args, has_labels=False, fgsm_eps, deep_args, has_labels=False, fgsm_eps, f
if device.type == 'cuda': torch.cuda.empty_cache()
```









robustness: 8.71e-04 eps: 0.001





true label : bird pred label : bird conf score : 16.76



pred label : bird conf score : 16.66



pred label : cat conf score : 11.04



robustness: 4.81e-04 eps: 0.001



robustness: 1.86e-02 overshoot: 0.02 iters: 4

Original

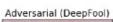


true label : cat pred label : cat conf score : 28.27

Adversarial (FGSM)



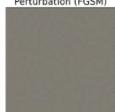
pred label : cat conf score : 23.79





pred label : dog conf score : 23.59

Perturbation (FGSM)



robustness: 1.83e-02 eps: 0.02



robustness: 1.37e-02 overshoot: 0.02 iters: 3



true label : horse pred label : horse conf score : 45.79



pred label : horse conf score : 41.30



pred label : deer conf score : 35.28



robustness: 1.70e-02 eps: 0.02



robustness: 3.21e-02 overshoot: 0.02 iters: 2

Original



true label : frog pred label : frog conf score : 34.21

Adversarial (FGSM)

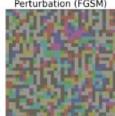


pred label : bird conf score : 18.70



pred label : bird conf score : 24.26

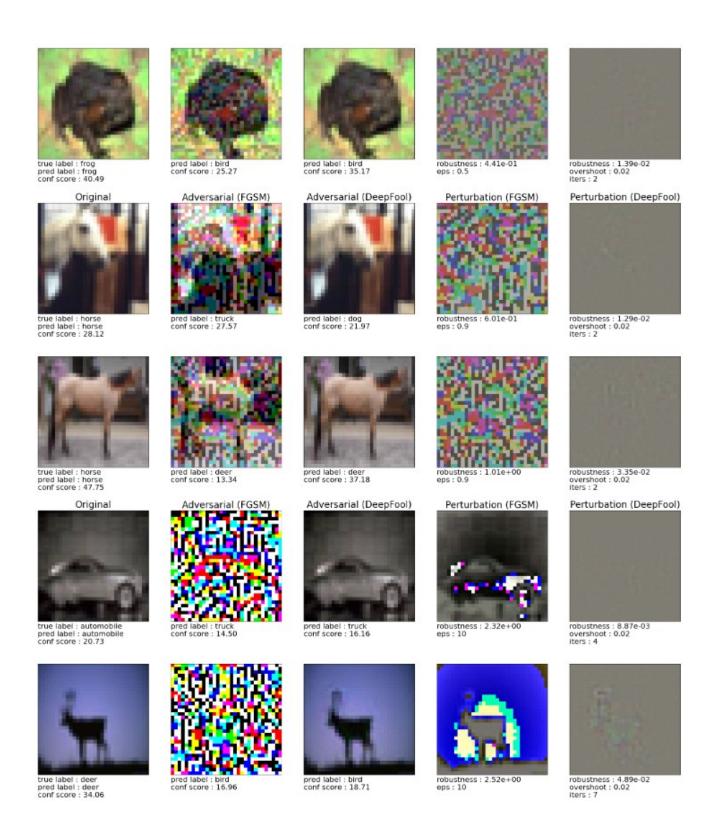
Perturbation (FGSM)



robustness: 7.10e-01 eps: 0.5



robustness : 2.65e-02 overshoot : 0.02 iters : 2



• Проверим влияние параметра fgsm_esp для LeNet на датасете Cifar-10.

```
fgsm_eps = 0.001
 model = LeNet_CIFAR().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_lenet.pth'))
display_attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_labels=False, 12_nometric model.
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm eps = 0.02
model = LeNet_CIFAR().to(device)
 model.load_state_dict(torch.load('weights/clean/cifar_lenet.pth'))
display_attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_labels=False, 12_noi
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm eps = 0.5
model = LeNet_CIFAR().to(device)
 model.load_state_dict(torch.load('weights/clean/cifar_lenet.pth'))
display_attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_labels=False, 12_nometric files and files are files are files and files are files a
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 0.9
model = LeNet_CIFAR().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_lenet.pth'))
display_attack(device, model, cifar_test, cifar_tf_inv, cifar_min, cifar_max, fgsm_eps, deep_args, has_labels=False, 12_nor
if device.type == 'cuda': torch.cuda.empty_cache()
fgsm_eps = 10
model = LeNet_CIFAR().to(device)
model.load_state_dict(torch.load('weights/clean/cifar_lenet.pth'))
\label{lem:display_attack} device, \ model, \ cifar\_test, \ cifar\_tf\_inv, \ cifar\_min, \ cifar\_max, \ fgsm\_eps, \ deep\_args, \ has\_labels= \textit{False}, \ 12\_noinder \ fgsm\_eps, \ deep\_args, \ has\_labels= \textit{False}, \ 12\_noinder \ fgsm\_eps, \ deep\_args, \ has\_labels= \textit{False}, \ 12\_noinder \ fgsm\_eps, \ deep\_args, \ has\_labels= \textit{False}, \ 12\_noinder \ fgsm\_eps, \ deep\_args, \ has\_labels= \textit{False}, \ 12\_noinder \ fgsm\_eps, \ deep\_args, \ has\_labels= \textit{False}, \ 12\_noinder \ fgsm\_eps, \ deep\_args, \ has\_labels= \textit{False}, \ 12\_noinder \ fgsm\_eps, \ has\_labels= \ has\_labels= 
if device.type == 'cuda': torch.cuda.empty_cache()
```



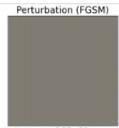
true label : deer pred label : deer conf score : 8.35



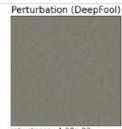
pred label : deer conf score : 8.29



pred label : horse



robustness: 8.19e-04 eps: 0.001



robustness: 1.60e-02 overshoot: 0.02 iters: 3



true label : truck pred label : automobile conf score : 7.19



pred label : automobile conf score : 7.32



pred label : ship conf score : 6.05





robustness: 4.48e-03 overshoot: 0.02 iters: 2

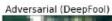


true label : horse pred label : horse conf score : 24.45



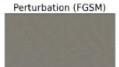


pred label : horse conf score : 21.08





pred label : dog conf score : 13.60



robustness: 1.83e-02 eps: 0.02



robustness : 4.95e-02 overshoot : 0.02 iters : 3



true label : truck pred label : cat conf score : 2.91



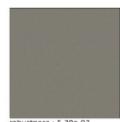
pred label : cat conf score : 3.28



pred label : dog conf score : 2.61



robustness : 1.99e-02 eps : 0.02



robustness: 5.30e-03 overshoot: 0.02 iters: 3



true label : ship pred label : ship conf score : 17.76

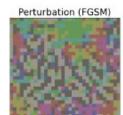
Adversarial (FGSM)



pred label : bird conf score : 4.53



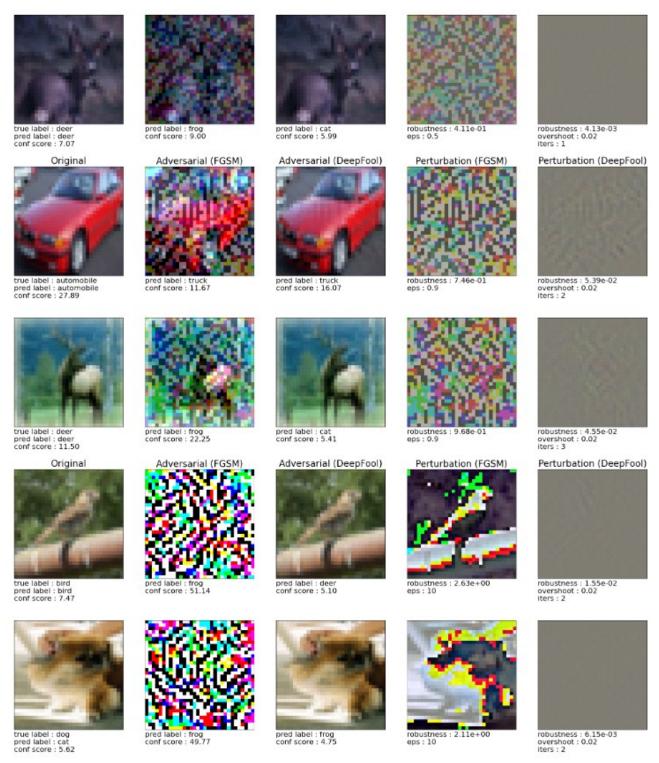
pred label : airplane conf score : 10.46



robustness: 4.61e-01 eps: 0.5



robustness : 3.11e-02 overshoot : 0.02 iters : 2



Вывод: параметр fgsm_esp влияет на устойчивость сети. При увеличении значения fgsm_esp сети становятся уязвимыми к атакам и увеличивается ошибка тестирования модели. Это указывает на снижение ее производительности