



Attacks on AI made easy

Elizaveta Tishina

Security Analysis Specialist, DeteAct

Moscow, August 26, 2022

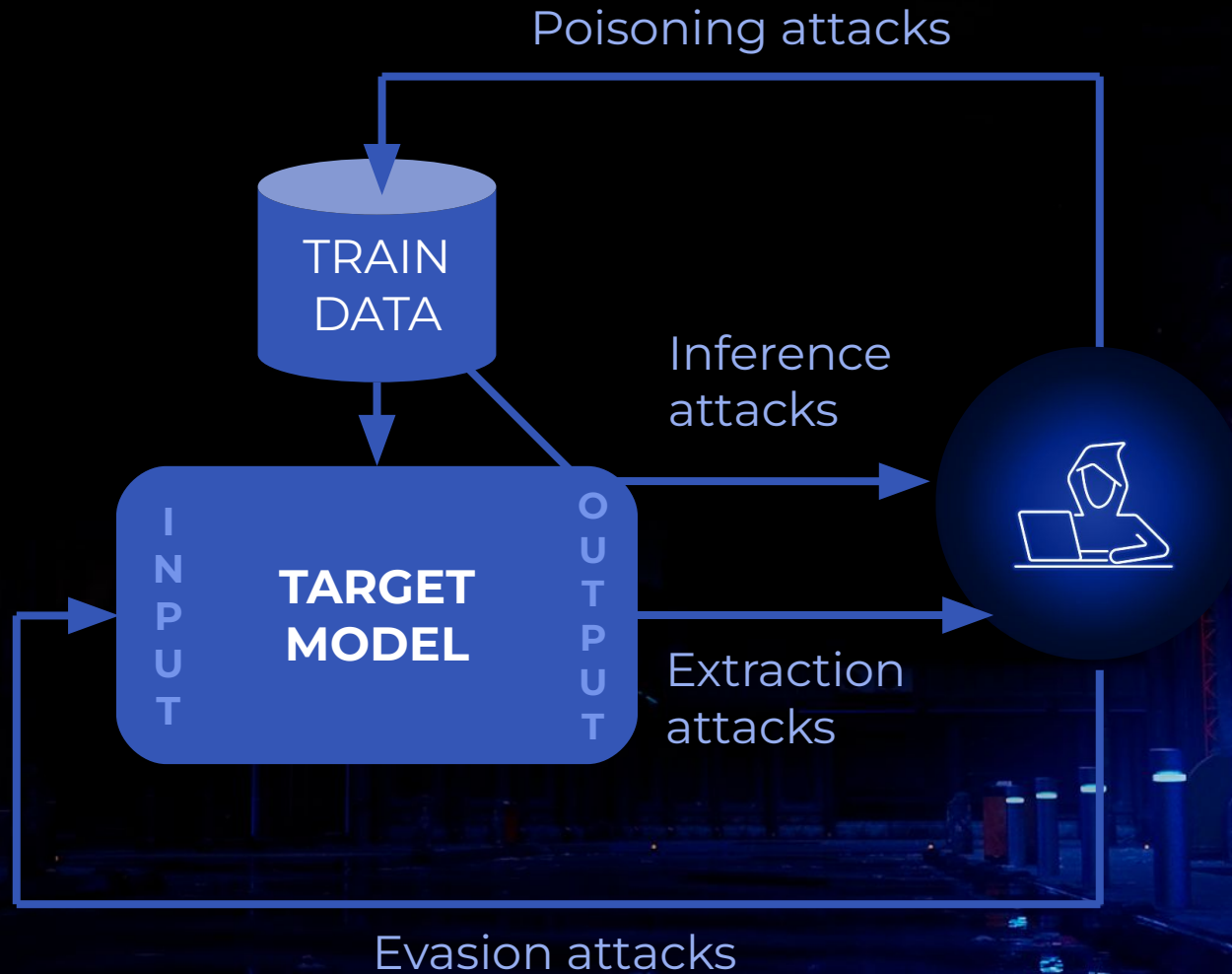


Machine Learning in products

- Recommendation systems
- Identity verification
- Malware detection
- Spam detection
- Search engines
- Translation
- Machine vision
- Fraud detection
- Analysis of results of medical tests

...

Threats to Machine Learning models

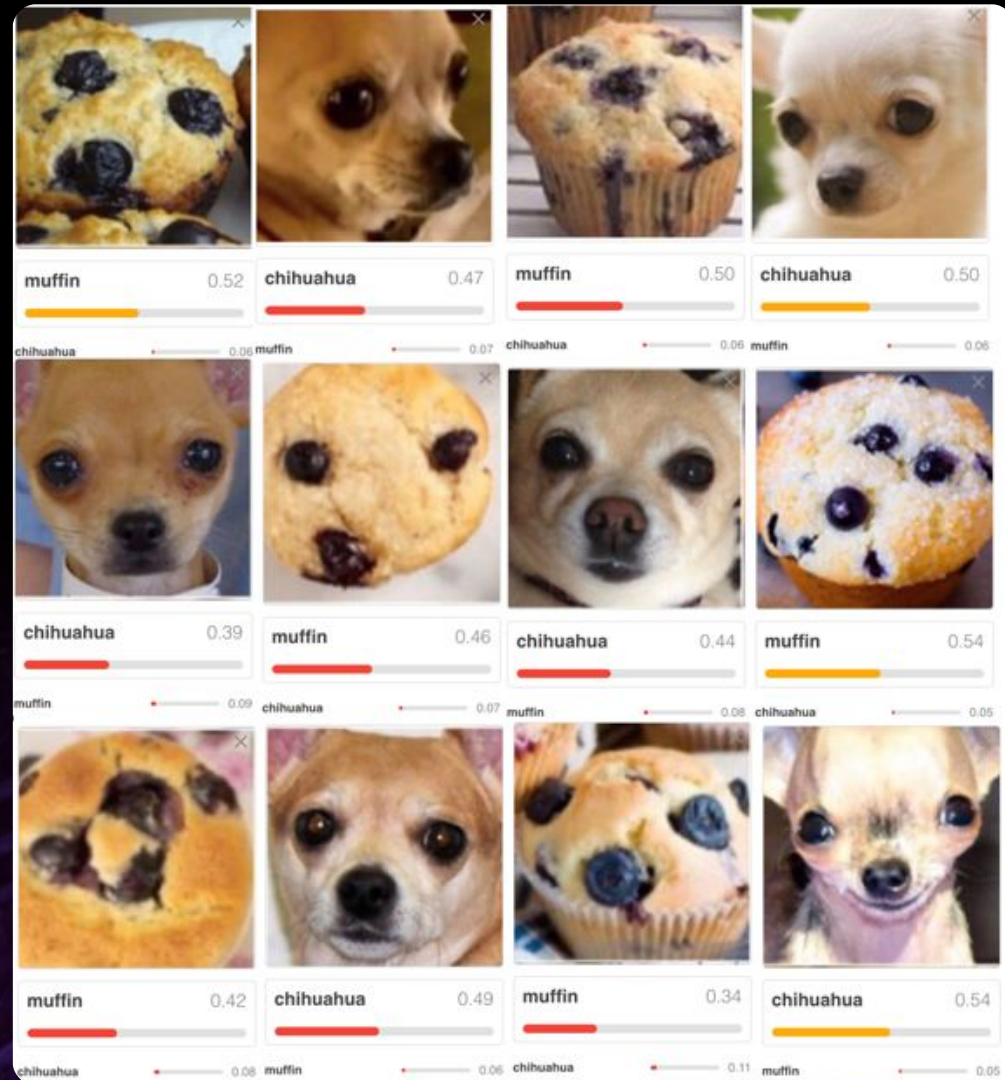


Example model overview

- **Train set:** MNIST dataset
 - **Prerequisites:** API access to model
 - **Attacker knowledge:** –
- Forms processing
 - Bank cheques processing

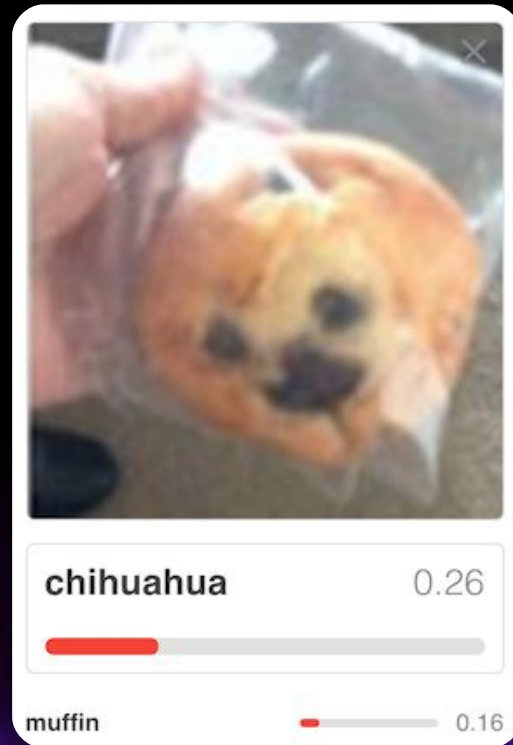
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9
0	1	2	3	4	5	6	7	8	9

Evasion attack: Overview

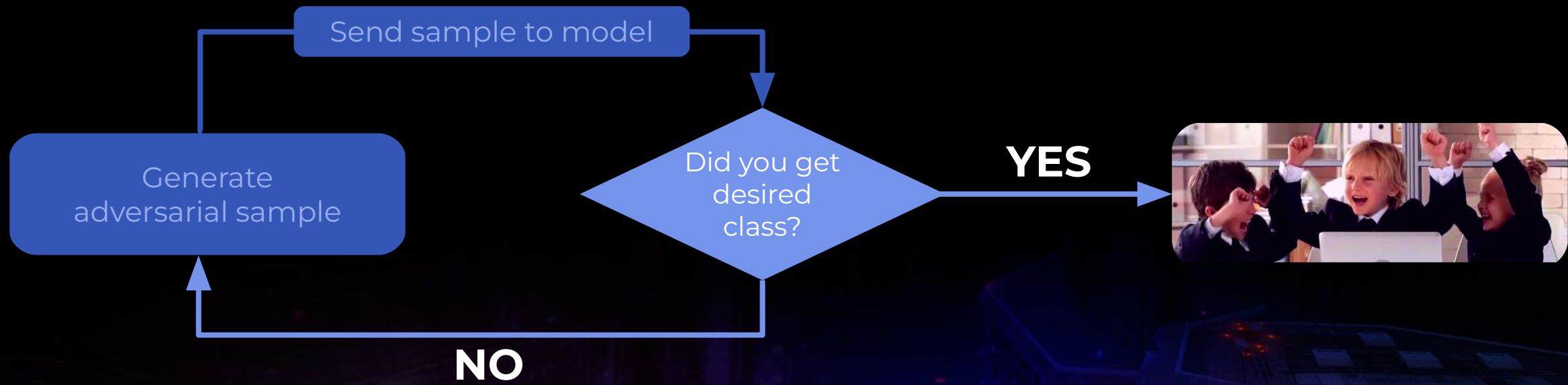


OFFZONE

Evasion attack: Overview

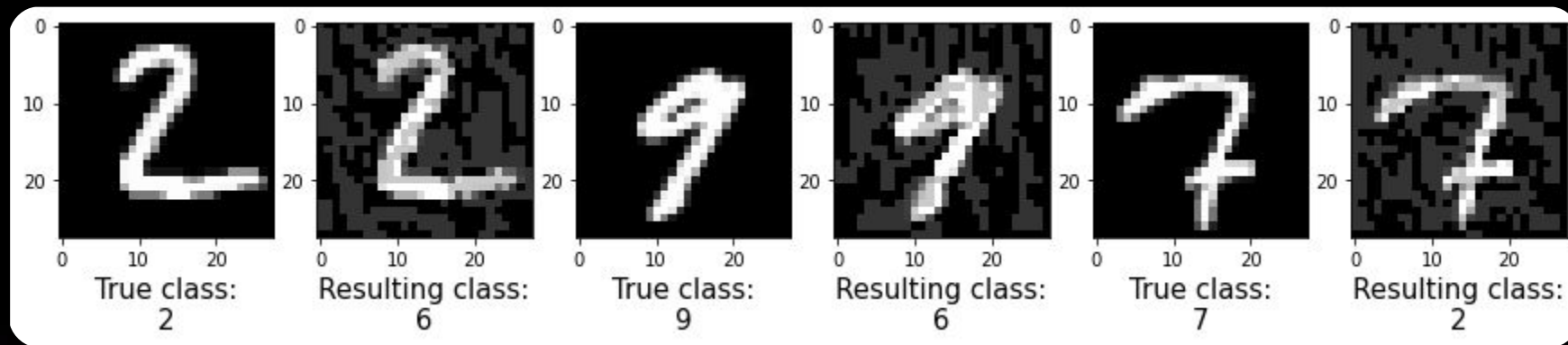


Evasion attack: Overview



Evasion attack: Example

 github.com/qwqoro/ML-Talk

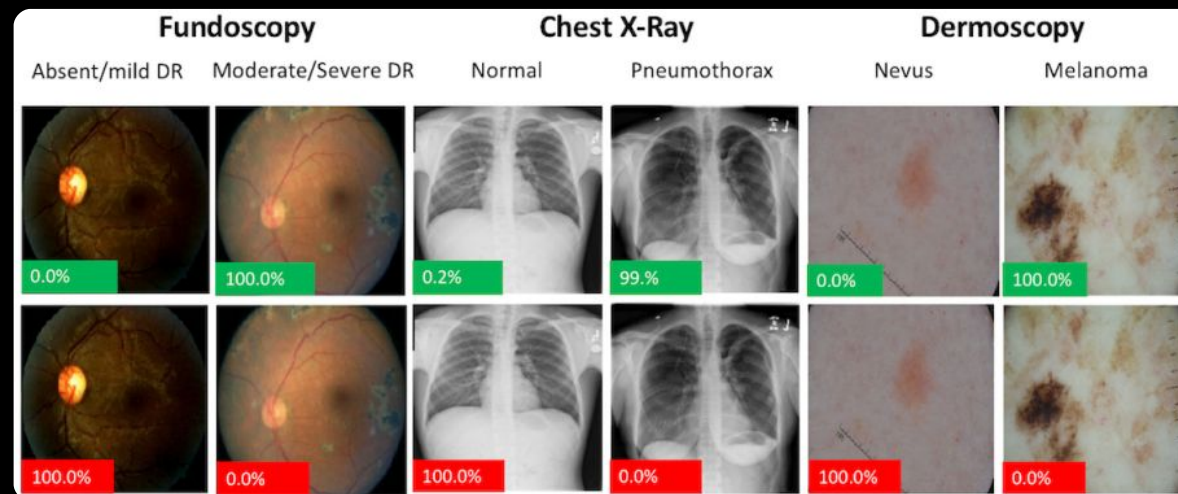


Approach: Fast Gradient Method [arXiv:1412.6572]



Evasion attack: Impact

- Malware filters bypass / Antivirus evasion
- Spam e-mails & ad filters bypass
- Spoofing against verification systems
- Life-threatening situations

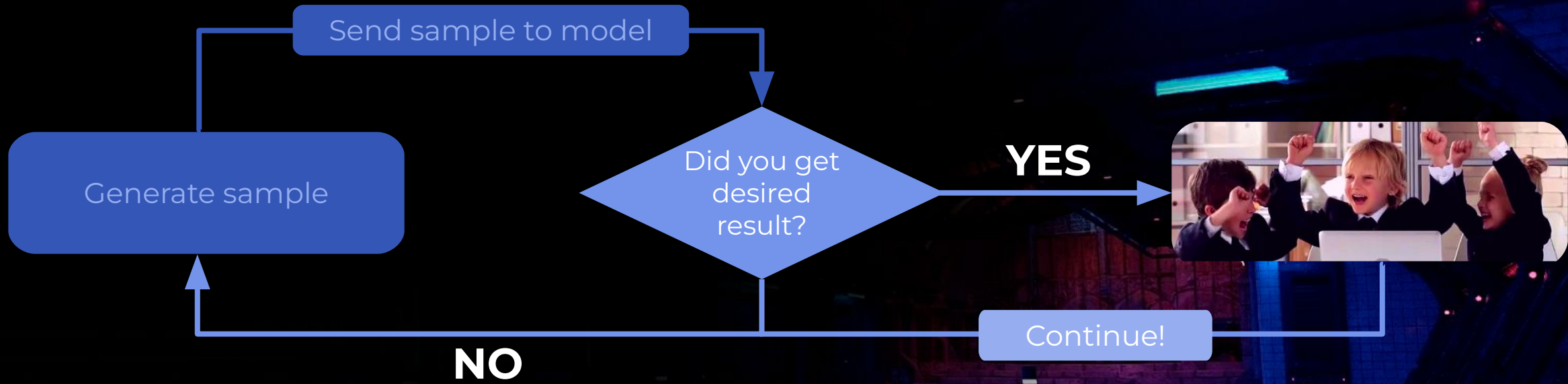


[arXiv:1804.05296]

From Attacker <attacker@bk.ru> ☆	From Attacker <attacker@bk.ru> ☆
Subject [REDACTED]	Subject [REDACTED]
To Victim@victim.com ☆	To Victim@victim.com ☆
Hey there	Heythere
So I'm the hacker who cracked your e mail	SoMl'mgthemhackerNwhohcrackedEyourbeFmail
You entered your password on one of the sites	YouNenterediyourjpasswordRonconetofAthelsites
This is the password from	Thisisrtherpasswordyfromxjevans@pearlholding
Obviously you can can change it, or perhaps	ObviouslyDyouLcanycanFchangevit_Zorlperhapsk

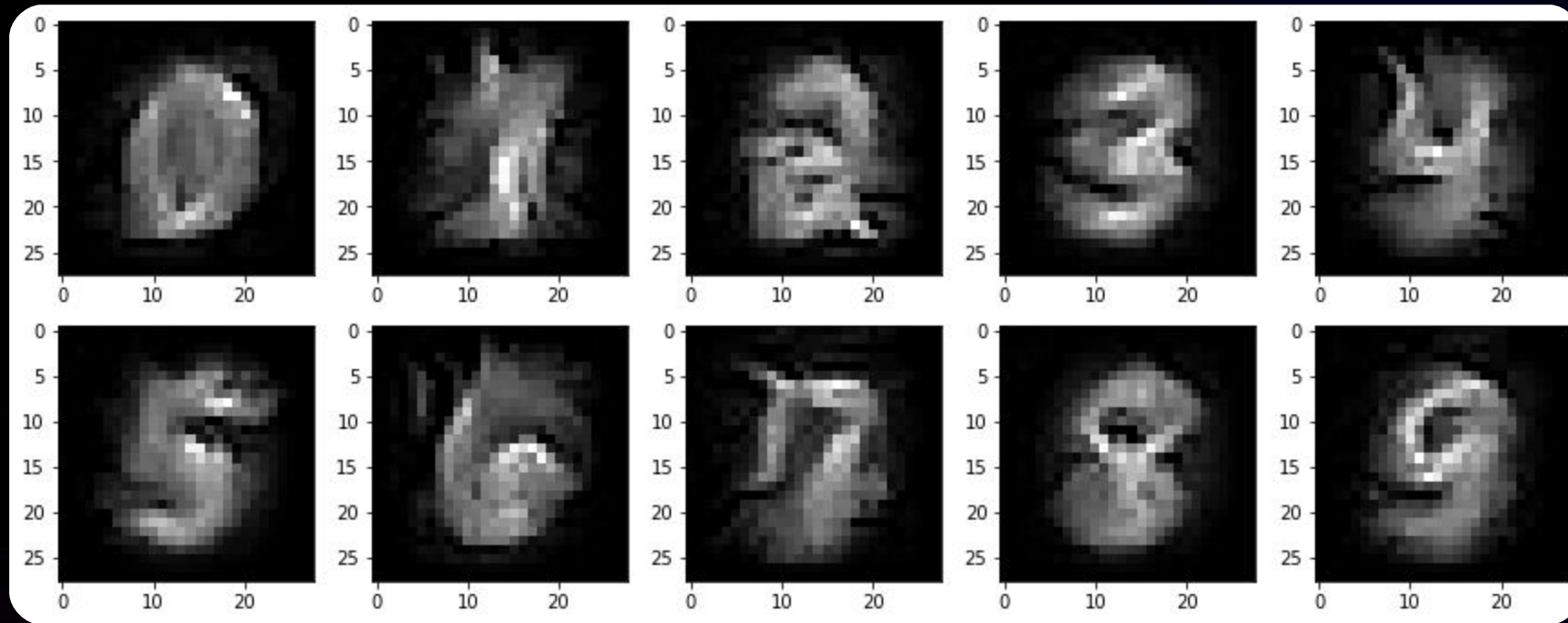


Model inversion attack: Overview



Model inversion attack: Example

 github.com/qwqoro/ML-Talk

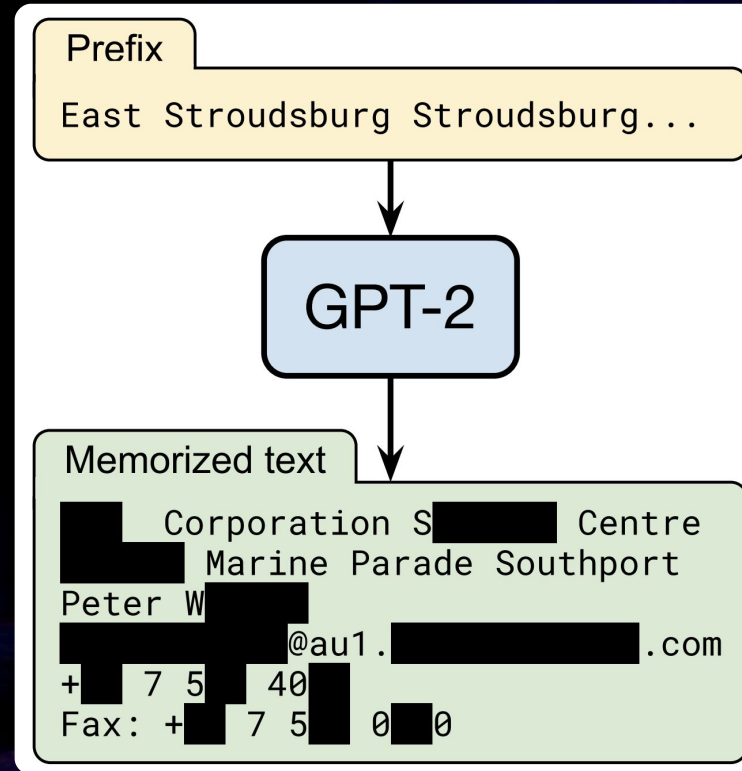


Approach: MIFace [[DOI:10.1145/2810103.2813677](https://doi.org/10.1145/2810103.2813677)]

Model inversion attack: Impact

Leak of sensitive data:

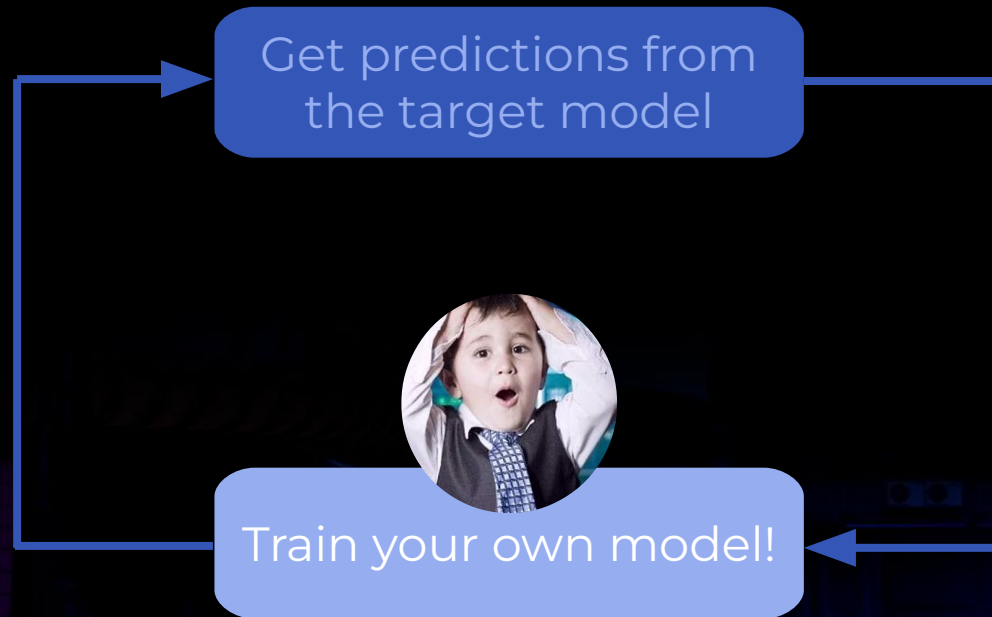
- Contents of documents
- Medical records
- Passwords
- PIN codes
- Other secrets



[arXiv:2012.07805]

Model extraction attack: Overview & Impact

- Intellectual property infringement



Adversarial Robustness Toolbox

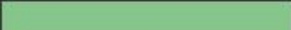
 github.com/Trusted-AI/adversarial-robustness-toolbox

- Attacks
- Defences
- Estimators
- Metrics
- Data generators
- Examples & detailed notebooks

```
from art.attacks.inference.model_inversion import MIFace

x_average = np.zeros((10, 28, 28, 1)) + np.mean(x_test, axis=0)

attackInversion = MIFace(classifier, max_iter=25000, threshold=1.0, batch_size=10, window_length=128)
inverted = attackInversion.infer(x_average, y=np.arange(10))
```

Model inversion: 100%  10/10 [10:13<00:00, 61.38s/it]

```
from art.attacks.evasion import FastGradientMethod

# Generation of adversarial examples
attackEvasion = FastGradientMethod(estimator=classifier, eps=0.2, batch_size=64)
x_adv = attackEvasion.generate(x_test)

# Predicting and evaluating accuracies of predictions on both initial data samples and adversarial ones
predictions = (classifier.predict(x_test), classifier.predict(x_adv))
accuracies = (np.sum(np.argmax(predictions[0], axis=1) == np.argmax(y_test, axis=1)) / len(y_test),
              np.sum(np.argmax(predictions[1], axis=1) == np.argmax(y_test, axis=1)) / len(y_test))

print(f"Accuracy of predictions (initial data): {accuracies[0] * 100} %")
print(f"Accuracy of predictions (adversarial): {accuracies[1] * 100} %")
```

Accuracy of predictions (initial data): 98.16 %
Accuracy of predictions (adversarial): 41.88 %

[ART] Model extraction attack: Example

 github.com/qwqoro/ML-Talk

```
from art.attacks.extraction import CopycatCNN

# Training a substitute model based on the target model
attackExtraction = CopycatCNN(classifier, batch_size_fit=10, batch_size_query=10, nb_epochs=10, nb_stolen=100)
extracted = attackExtraction.extract(x_test, thieved_classifier=res)

Train on 100 samples
Epoch 1/10
100/100 [=====] - 0s 2ms/sample - loss: 2.2616 - accuracy: 0.1500
Epoch 2/10
100/100 [=====] - 0s 740us/sample - loss: 2.0509 - accuracy: 0.2600
Epoch 3/10
100/100 [=====] - 0s 658us/sample - loss: 1.7601 - accuracy: 0.4600
Epoch 4/10
100/100 [=====] - 0s 658us/sample - loss: 1.4344 - accuracy: 0.5800
Epoch 5/10
100/100 [=====] - 0s 702us/sample - loss: 1.1608 - accuracy: 0.6700
Epoch 6/10
100/100 [=====] - 0s 707us/sample - loss: 0.9168 - accuracy: 0.7400
Epoch 7/10
100/100 [=====] - 0s 804us/sample - loss: 0.7963 - accuracy: 0.7500
Epoch 8/10
100/100 [=====] - 0s 700us/sample - loss: 0.6875 - accuracy: 0.7900
Epoch 9/10
100/100 [=====] - 0s 570us/sample - loss: 0.6211 - accuracy: 0.8100
Epoch 10/10
100/100 [=====] - 0s 619us/sample - loss: 0.5331 - accuracy: 0.8100

# Making predictions with use of both original and extracted versions of the target model and evaluating their similarity
victim_predictions = np.argmax(model.predict(x_test), axis=1)
thieved_predictions = np.argmax(extracted.predict(x_test), axis=1)
accuracy = np.sum(victim_predictions == thieved_predictions) / len(victim_predictions)

print(f"Similarity of predictions: {accuracy * 100} %")

Similarity of predictions: 69.31 %
```

Counterfit

 github.com/Azure/counterfit

```
/content/counterfit# python counterfit.py
```



Version: 1.0.0

```
counterfit> list targets
```

Name	Model Type	Data Type	Input Shape	# Samples	Endpoint	Loaded
creditfraud	BlackBox	tabular	(30,)	(not loaded)	creditfraud_sklearn_pipeline.pkl	False
digits_blackbox	BlackBox	image	(1, 28, 28)	(not loaded)	mnist_sklearn_pipeline.pkl	False
digits_keras	keras	image	(28, 28, 1)	(not loaded)	mnist_model.h5	False
movie_reviews	BlackBox	text	(1,)	(not loaded)	movie_reviews_sentiment_analysis.pt	False
satellite	BlackBox	image	(3, 256, 256)	(not loaded)	satellite-image-params-airplane-stadium.h5	False

```
counterfit> list frameworks
```

Framework	# Attacks
art	(not loaded)
augly	(not loaded)
textattack	(not loaded)

Counterfit

 github.com/Azure/counterfit

counterfit> list attacks

Name	Category	Type	Tags	Framework					
A2TYoo2021	BlackBox	EvasionAttack	text		WhiteBox	EvasionAttack	image, tabular	art	
BAEGarg2019	BlackBox	EvasionAttack	text		WhiteBox	EvasionAttack	image, tabular	art	
BERTAttackLi2020	BlackBox	EvasionAttack	text		BlackBox	ExtractionAttack	image	art	
CLARE2020	BlackBox	EvasionAttack	text	CarliniL0Method	WhiteBox	EvasionAttack	image, tabular	art	
CheckList2020	BlackBox	EvasionAttack	text	CarliniLInfMethod	WhiteBox	EvasionAttack	image, tabular	art	
DeepWordBugGao2018	BlackBox	EvasionAttack	text	CopycatCNN	BlackBox	ExtractionAttack	image	art	
FasterGeneticAlgorithmJia2019	BlackBox	EvasionAttack	text	DeepFool	WhiteBox	EvasionAttack	image, tabular	art	
GeneticAlgorithmAlzantot2018	BlackBox	EvasionAttack	text	ElasticNet	WhiteBox	EvasionAttack	image, tabular	art	
HotFlipEbrahimi2017	BlackBox	EvasionAttack	text	FunctionallyEquivalentExtraction	BlackBox	ExtractionAttack	image, tabular	art	
IGAWang2019	BlackBox	EvasionAttack	text	HopSkipJump	BlackBox	EvasionAttack	image, tabular	art	
InputReductionFeng2018	BlackBox	EvasionAttack	text	KnockoffNets	BlackBox	ExtractionAttack	image, tabular	art	
Kuleshov2017	BlackBox	EvasionAttack	text	LabelOnlyDecisionBoundary	WhiteBox	InferenceAttack	image, tabular	art	
MorpheusTan2020	BlackBox	EvasionAttack	text	MIFace	WhiteBox	InferenceAttack	image, tabular	art	
PSOZang2020	BlackBox	EvasionAttack	text	NewtonFool	WhiteBox	EvasionAttack	image, tabular	art	
PWWSRen2019	BlackBox	EvasionAttack	text	ProjectedGradientDescentCommon	WhiteBox	EvasionAttack	image, tabular	art	
Pruthi2019	BlackBox	EvasionAttack	text	SaliencyMapMethod	WhiteBox	EvasionAttack	image, tabular	art	
Seq2SickCheng2018BlackBox	BlackBox	IntegrityAttack	text	SimBA	WhiteBox	EvasionAttack	image	art	
TextBuggerLi2018	BlackBox	EvasionAttack	text	SpatialTransformation	WhiteBox	EvasionAttack	image, tabular	art	
TextFoolerJin2019	BlackBox	EvasionAttack	text	UniversalPerturbation	WhiteBox	EvasionAttack	image	art	
BoundaryAttack	BlackBox	EvasionAttack	image	VirtualAdversarialMethod	WhiteBox	EvasionAttack	image	art	
				Wasserstein	WhiteBox	EvasionAttack	image	art	
				Blur	BlackBox	CommonCorruption	image	augly	
				Brightness	BlackBox	CommonCorruption	image	augly	
				ChangeAspectRatio	BlackBox	CommonCorruption	image	augly	
				ClipImageSize	BlackBox	CommonCorruption	image	augly	
				ColorJitter	BlackBox	CommonCorruption	image	augly	
				Contrast	BlackBox	CommonCorruption	image	augly	

[Counterfit] Evasion attack: Example

 github.com/qwqoro/ML-Talk

```
digits_blackbox> use HopSkipJump

[+] New HopSkipJump (419f7593) created
[+] Using 419f7593

digits_blackbox>419f7593> set --sample_index 1 --max_eval 1500 --max_iter 10
```

```
digits_blackbox>419f7593> run

[-] Running attack HopSkipJump with id 419f7593 on digits_blackbox)

[-] Preparing attack...
[-] Running attack...
```

Success	Elapsed time	Total Queries
1/1	0.7	2390 (3504.1 query/sec)










Sample Index	Input Label (conf)	Adversari... Label (conf)	Max Abs Chg.	Adversarial Input
1	0 (1.0000)	6 (0.9809)	4.7776	counterfit/targets/digits_blackbox/results/419f7593/digits_blackbox-f03b8b22-f

```
[+] Attack completed 419f7593 (HopSkipJump)
```

Approach: HopSkipJump [arXiv:1904.02144]

[Counterfit] Evasion attack: Example

 github.com/qwqoro/ML-Talk

Original image	Adversarial version of image	Difference (exaggerated)	Original class	Original confidence	Resulting class	Resulting confidence
			0	100%	6	98%
			1	100%	8	56%
			2	100%	4	73%

Counterfit

 github.com/Azure/counterfit

- Hiding malicious queries
- Using a proxy
- Sending and collecting outputs from different locations
- Adding startup commands
- Overriding functions in the parent target
- Training a local model to attack

NO
FF
ONE
2022

DETECT

 @qwqoro

 qwqoro/ML-Talk





**NO
OFF
ONE
2022**