



**SERMA**  
SAFETY & SECURITY

# Improving Leakage Exploitability in Horizontal Side Channel Attacks through Anomaly Mitigation with Unsupervised Neural Networks

Gauthier Cler, Sébastien Ordas, Philippe Maurine



# Outline

- 1 Horizontal Attacks
- 2 Impact of anomalies on Pol selection
- 3 Anomalies mitigation
- 4 Results
- 5 Conclusion

# Horizontal Attacks

# Horizontal Attacks

- ▶ Single trace attack
- ▶ No profiling on open device possible, no leakage assessment.
- ▶ Usually on asymmetric implementations (RSA, ECC).
- ▶ Clustering approach:
  - 1 Divide trace into patterns
  - 2 Points of Interest (PoI) selection with univariate clustering
  - 3 Multidimensional clustering

**Attack success highly relies on the quality of the trace.**

## Impact of anomalies on Pol selection

# Anomalies in data

Outliers (interquantile range)

Distribution tails

$$x \notin [Q_1 - 1.5 \times IQR, Q_3 + 1.5 \times IQR]$$

# Anomalies in data

Outliers (interquantile range)

Distribution tails

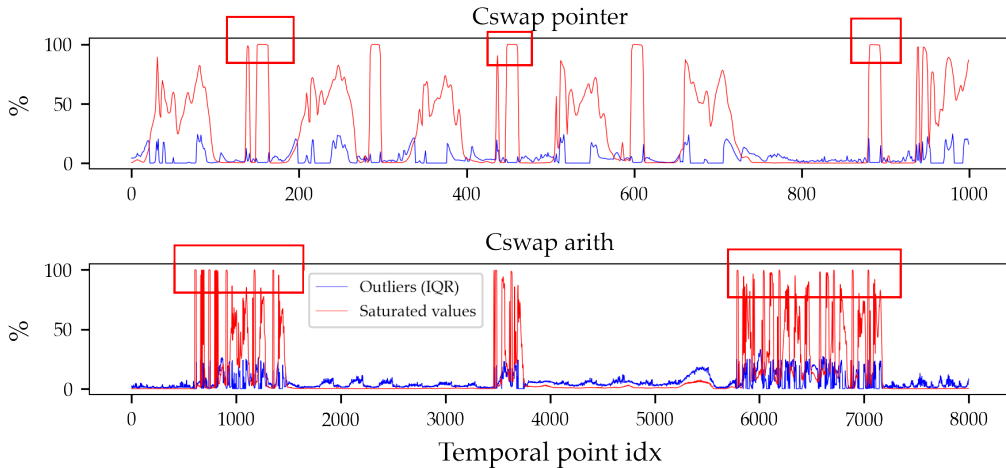
$$x \notin [Q_1 - 1.5 \times IQR, Q_3 + 1.5 \times IQR]$$

Saturated values

min/max values of digital sampling, for 8bit:

$$x = -128 \vee x = 127$$

# Anomalies in data

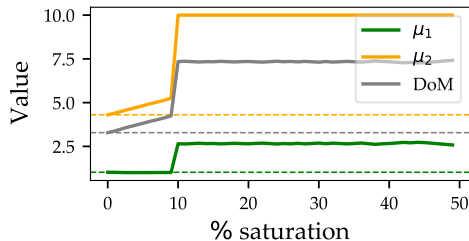
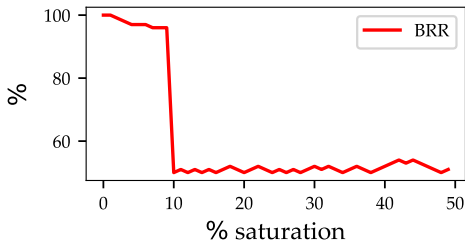


<sup>1</sup>Average anomalies Pointer:33.3%, Arith:16.5%



# Impact of anomalies on Pol selection

- ▶ Clustering is **not robust** to anomalies in data
- ▶ Can cause centroids shift, singularities,...



# Anomalies mitigation

# Limits of simple mitigation

## Mitigation by ablation

- ▶ Remove time points based on anomalies threshold
- ▶ Possibly losing information about the leakage

# Limits of simple mitigation

## Mitigation by ablation

- ▶ Remove time points based on anomalies threshold
- ▶ Possibly losing information about the leakage

## Mitigation by replacement

- ▶ Replace anomalies points with mean/median of non anomalies for each time point
- ▶ Decrease separability of mixture components

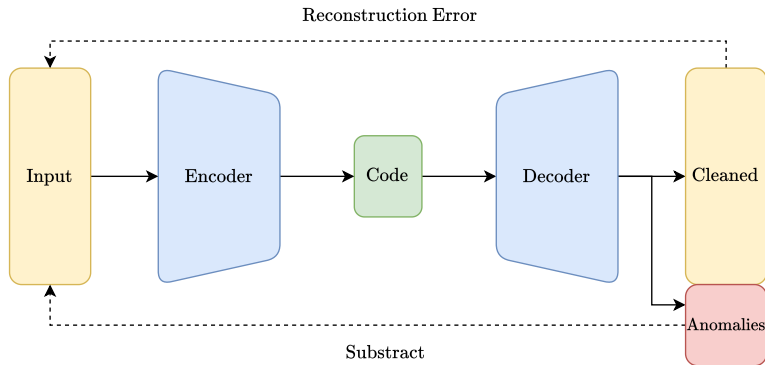
# Contribution - Mitigation with neural networks

Consider alternative methods

- ▶ Able to be trained in an unsupervised manner
- ▶ Leakage/information conservation
- ▶ Two approaches:
  - ⋮ Robust auto-encoder
  - ⋮ CycleGAN

# Robust auto-encoder unsupervised mitigation

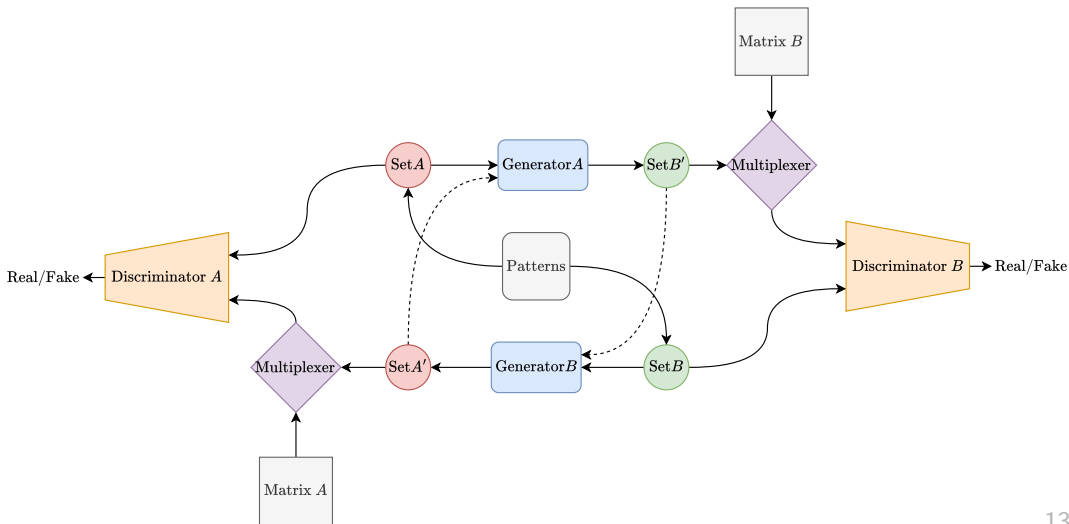
Decomposition of input data to **cleaned** and **anomalies** matrices.



# Limits

- ▶ RAE Generate new synthetic patterns → Can cause side effects on non anomalies.
- ▶ RAE does not exploit the anomalies model.

# Multiplexer CycleGAN self-supervised mitigation

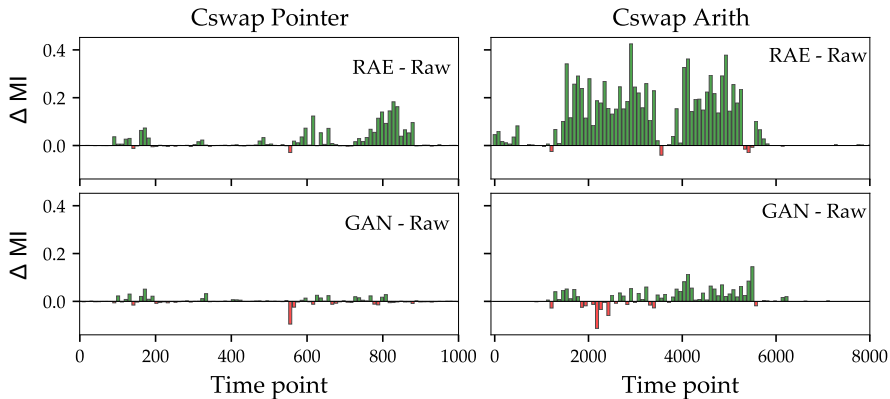




# Results

# Information conservation

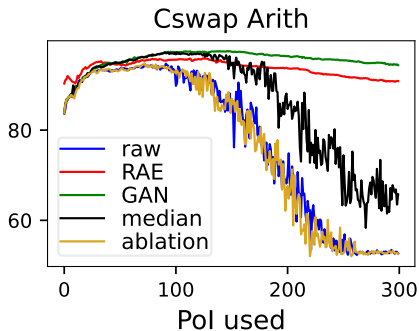
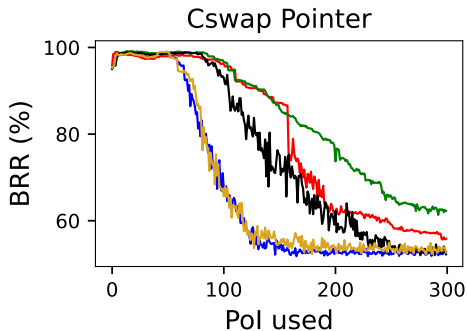
No change in the global MI. <sup>1</sup>



<sup>1</sup>Estimated with MINE.

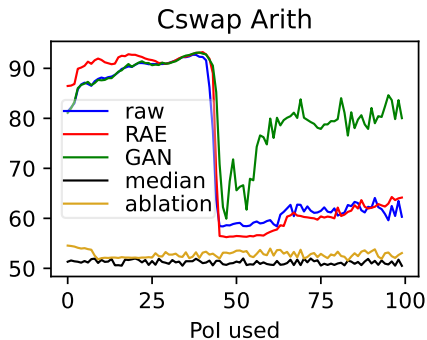
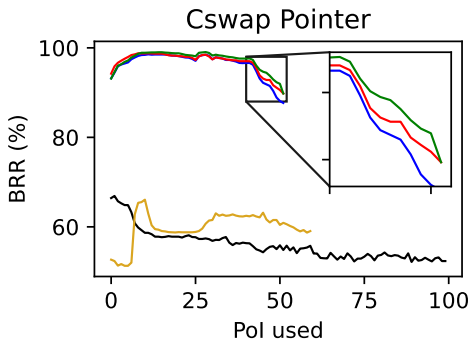
# Supervised selection - upper bound

Select  $k$  Pol with highest  $t$ -values and apply multidimensional clustering.



# Unsupervised selection

Multidimensional clustering on the best  $k$  Pol from Cler *et al.* 2023 unsupervised selection.



# Conclusion

# Conclusion

## Benefits

- ▶ Anomalies mitigation **improves leakage exploitability**
- ▶ Methods are applicable in a completely unsupervised context

# Conclusion

## Benefits

- ▶ Anomalies mitigation **improves leakage exploitability**
- ▶ Methods are applicable in a completely unsupervised context

## Limitations

- ▶ Architecture choice and parameters tuning can be hard in practice
- ▶ Attack success **still** depends on the exploitation method

# Conclusion

## Benefits

- ▶ Anomalies mitigation **improves leakage exploitability**
- ▶ Methods are applicable in a completely unsupervised context

## Limitations

- ▶ Architecture choice and parameters tuning can be hard in practice
- ▶ Attack success **still** depends on the exploitation method

## Future work

- ▶ Consider additional anomalies models
- ▶ Generalize on other targets/algorithms



Thank you for your attention.

Do you have any question?



**SERMA**

SAFETY & SECURITY

14, rue Galilée  
33600 PESSAC

05 57 26 08 88

[contact-s3@serma.com](mailto:contact-s3@serma.com)

# Bonus