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# **Little or Large?**

#### The effects of network size on AI explainability in Side-Channel Attacks

# 1. Knowledge Gap

#### Side-Channel Attacks

Side-Channel attacks are when a person or organization obtains information through the implementation of a computer system.

This work specifically looks at how encryption keys can be collected by observing power consumption during the encryption process

### **Research Question**

In the literature, the benefits of reducing network complexity are mentioned. However, no evidence that backs this claim was found.

This lead us to ask: what are the effects of network size on AI explainability?

## 2. Experiment

#### Visualization

relation

## Dataset

ASCAD:

#### Heatmap:

- Input/output
- Realistic •
- Frequently used
- Low computational overhead
  - **Architectures**

ZAID:

•

•

#### ASCAD:

- 5 blocks
- 1D convolution
- Average pooling
- Batch size 200
- Epochs 100 •
- Batch size 50 . •

1 block

Epochs 50

1D convolution

Average Pooling



The difference in area of interest likely stems from the different amount of convolutions are performed.

### Heatmap as function

It seems to be the case that neither of the two models perform better when a misclassification occurs.

The model large in complexity was shown to be inconsistent, indicating lower levels of explainability.

## Future work

The following topics show promise for future research:





## 4. Discussion

#### Heatmap overlay

### **Class level pattern**

# 5. Conclusion

#### Conclusions

A reduction in complexity leads to improved explainability because of:

- No notable changes in the misclassification patterns
  - Improvement in classification
  - explainability of the ZAID model
  - The effect different datasets have.
  - How different model behave
  - Output from different visualization techniques

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