

Double Strike: Breaking Approximation-Based Side-Channel Countermeasures for DNNs

SemSecuElec Seminar

Lorenzo **CASALINO**, Maria Mendez RÉAL, Jean-Christophe PRÉVOTET, Rubén SALVADOR

CentraleSupélec, IRISA, Inria (SUSHI Team)

`name.surname@irisa.fr`

Lab-STICC, Univ. Bretagne-Sud, UMR 6285

`maria.mendez-real@univ-ubs.fr`

Univ. Rennes, INSA Rennes, CNRS, IETR-UMR 6164

`jean-christophe.prevotet@insa-rennes.fr`

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Agenda

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- 2 Methodology
- 3 Results
- 4 Conclusions

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1 Background

2 Methodology

3 Results

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Deep Neural Networks (DNNs)

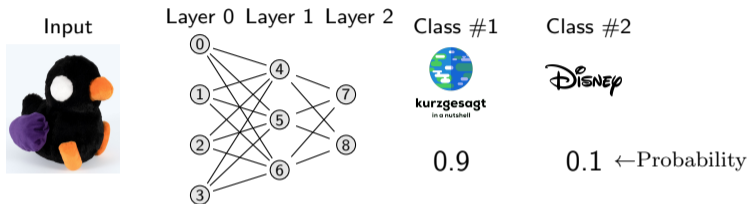
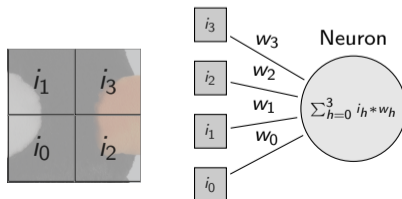


Figure: A DNN brand classifier¹.



¹Duck and Kurzgesagt Logo belong to Kurzgesagt

The *Weight(s)* of an Economic Damage

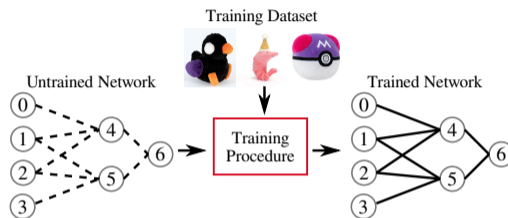


Figure: The training process.²

DNN Training is Expensive

- Expensive hardware (e.g., GPUs), time-intensive (e.g., days)

Weights Piracy

A non-negligible **economic damage**

²Duck (Kurzgesagt), Shrimp (Jellycat London), Masterball (Nintendo)

Side-Channel Analysis

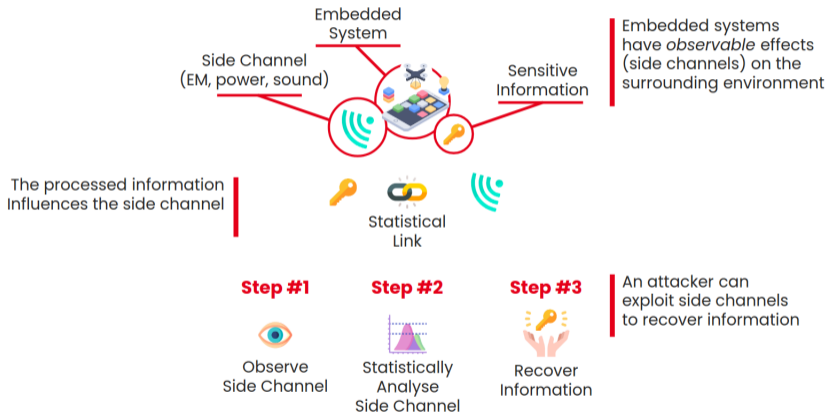


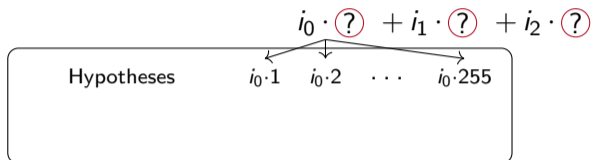
Figure: Side-Channel Analysis.

Side-channel Driven Weight Extraction

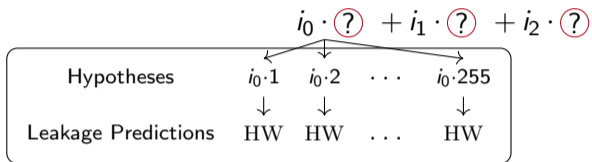
$$i_0 \cdot (?) + i_1 \cdot (?) + i_2 \cdot (?)$$

(↓)
the targets

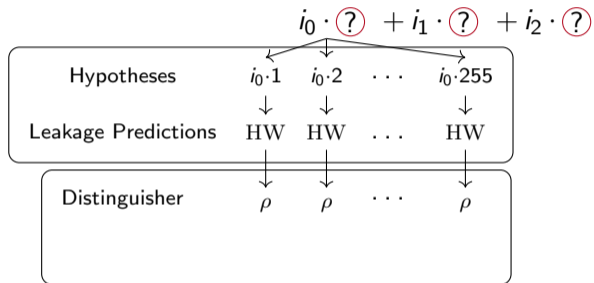
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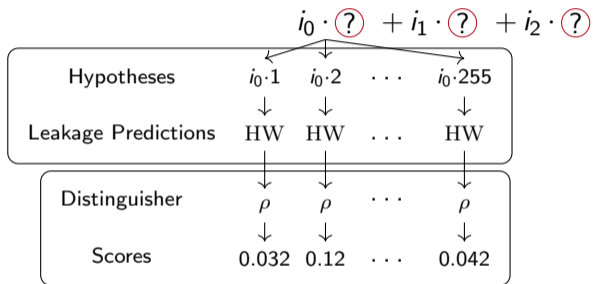
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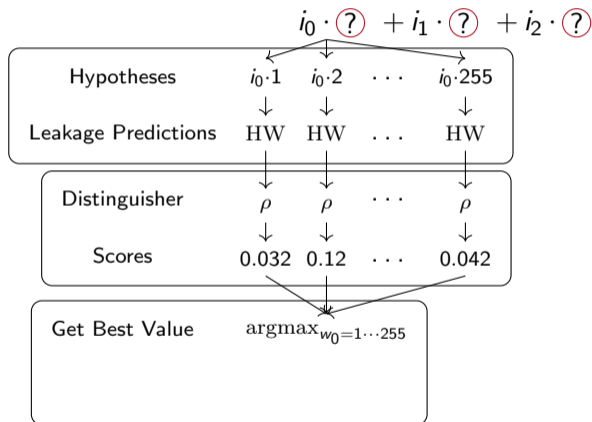
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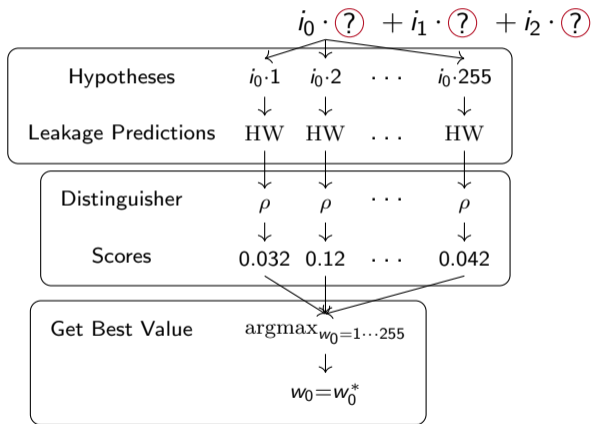
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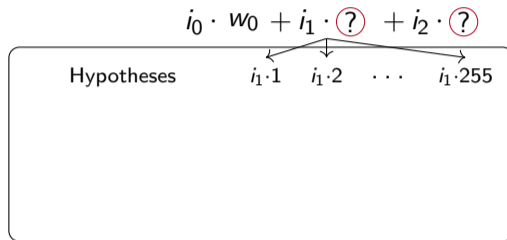
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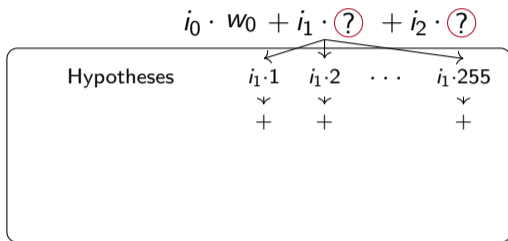
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$$i_0 \cdot w_0 + i_1 \cdot (?) + i_2 \cdot (?)$$

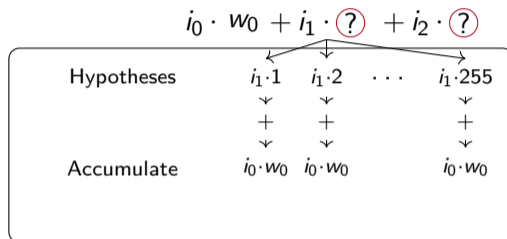
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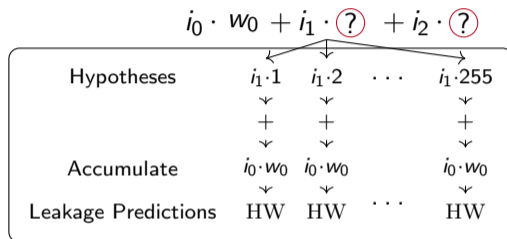
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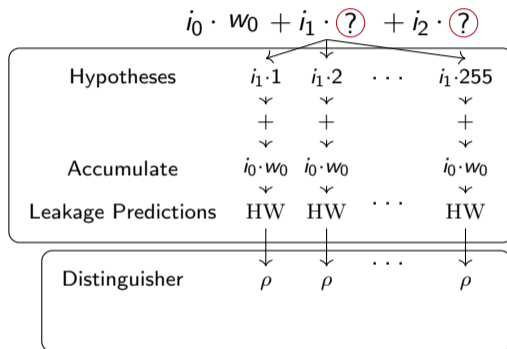
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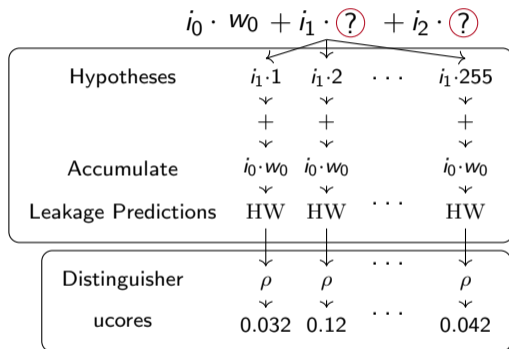
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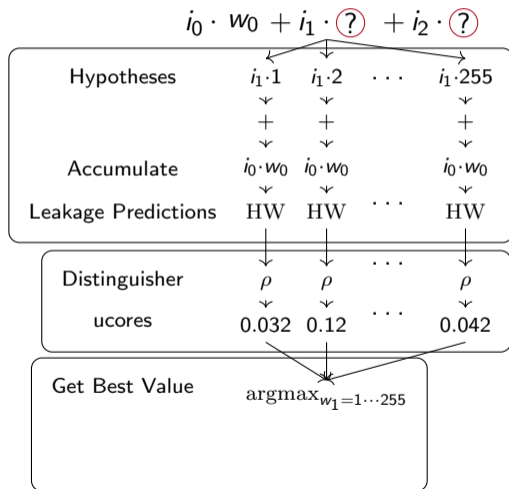
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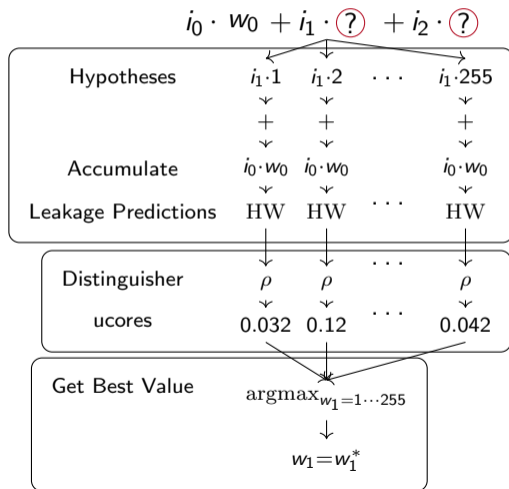
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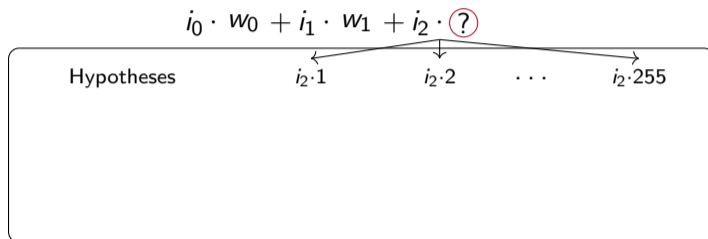
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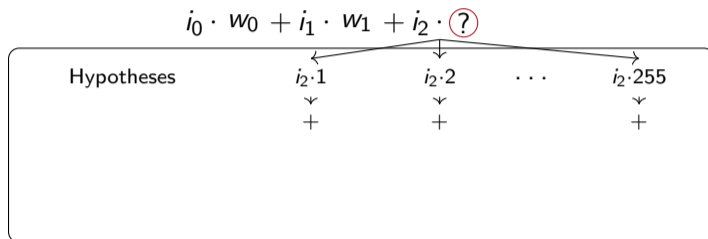
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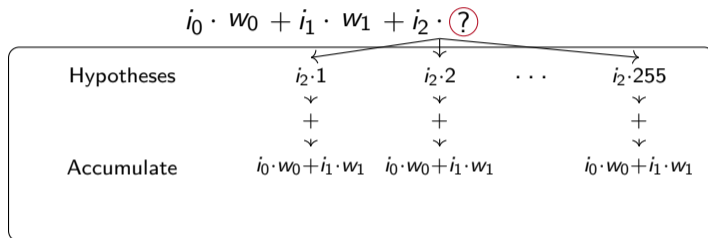
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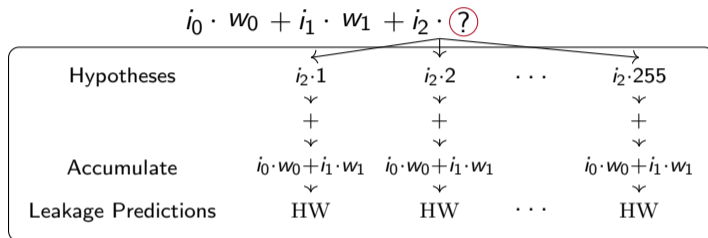
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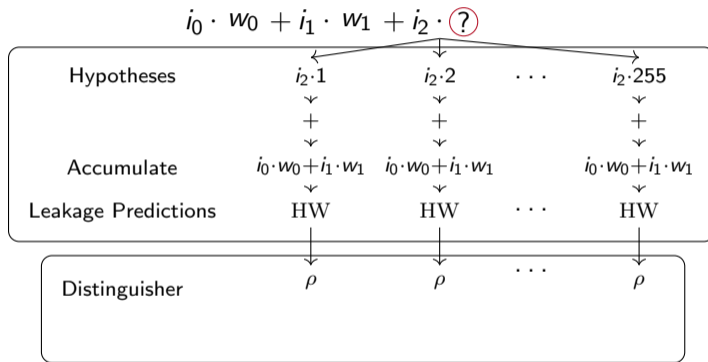
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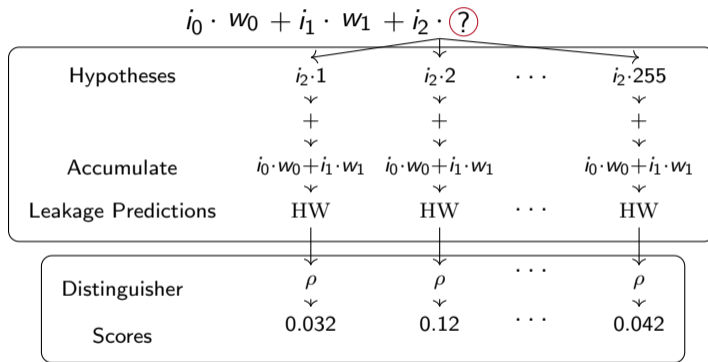
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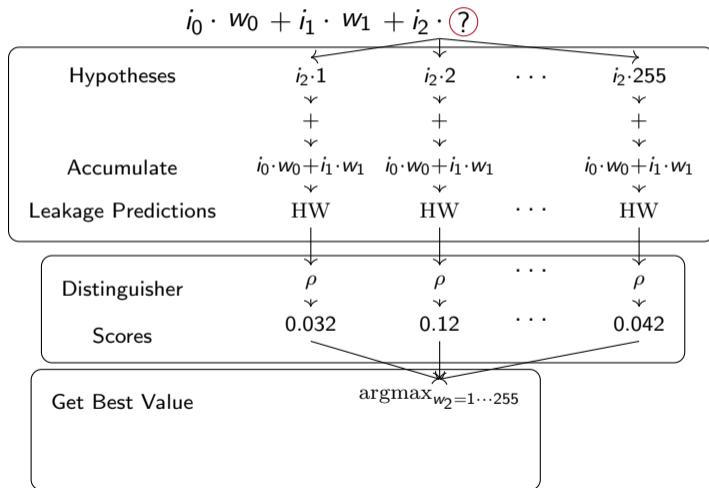
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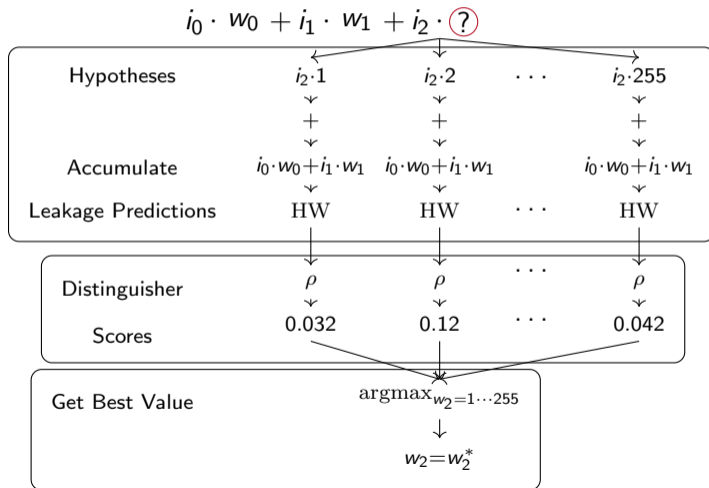
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Side-channel Driven Weight Extraction

$$i_0 \cdot w_0 + i_1 \cdot w_1 + i_2 \cdot w_2$$

Countermeasures

Masking

Replace the weight-dependent signal with N random ones

Shuffling

Randomly shuffle operations to bury weight-dependent signal in signal noise

DNN-Tailored Countermeasures

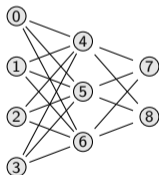
- Current defences come from cryptanalysis
- But DNNs \neq cryptosystems!
- DNNs exhibit particular characteristics (e.g., error resilience)

Approximate Computing

A Definition [Leo+25]

*Intentional insertion of errors to **trade tunable accuracy loss for valuable resource gains.***

Input



Class #1



0.9

Class #2



0.1

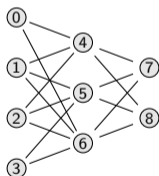
Original Network

Approximate Computing

A Definition [Leo+25]

*Intentional insertion of errors to **trade** tunable **accuracy** loss for valuable **resource** gains.*

Input



Class #1



0.7

Class #2



0.3

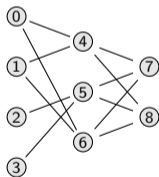
Pruned Network

Approximate Computing

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Input



Class #1



0.3

Class #2



0.7

Pruned Network

Approximate Computing and Side-Channel Behaviour

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Approximate-Computing (AxC)-based Countermeasures

- Approximation changes the application behaviour

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- Approximation changes the application's *side-channel signature*

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Approximate-Computing (AxC)-based Countermeasures

- Approximation changes the application behaviour
- Approximation changes the application's *side-channel signature*
- What role does AxC play in side-channel resilience? [Jap+25; Zha+26]

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 - Core question of the ANR "ATTILA" project

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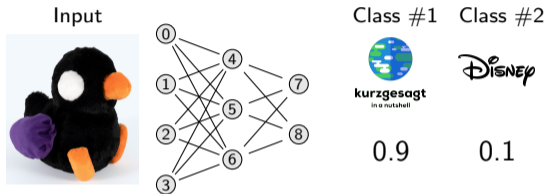
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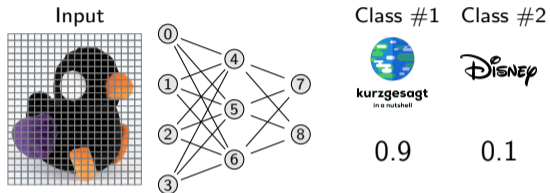
Approximate-Computing (AxC)-based Countermeasures

- Approximation changes the application behaviour
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- What role does AxC play in side-channel resilience? [Jap+25; Zha+26]
 - Core question of the ANR "ATTILA" project
- Recently considered as a side-channel countermeasure [Din+25]

Pixel Importance

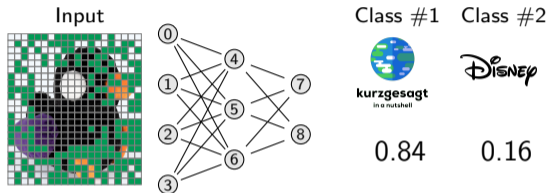


Pixel Importance



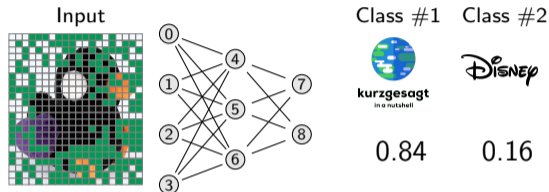
- What happens if we *prune* (skip) pixels?

Pixel Importance



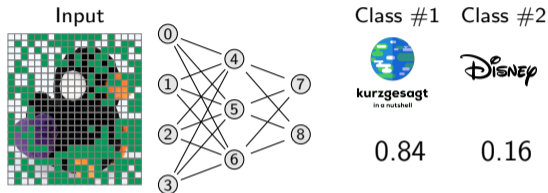
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- **Non-important pixel**: negligible accuracy reduction

Pixel Importance



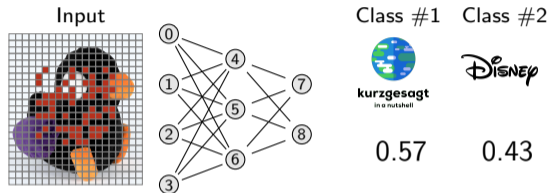
- What happens if we *prune* (skip) pixels?
- **Non-important pixel**: negligible accuracy reduction
 - **Non-important weight**

Pixel Importance



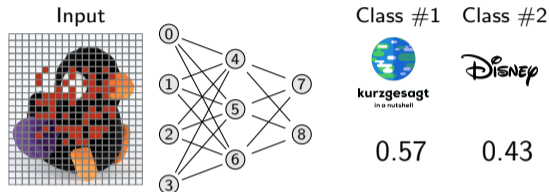
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 - Non-important Multiply-and-Accumulate (NIMAC)

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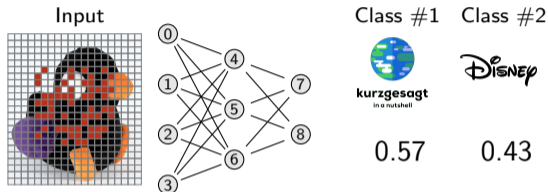
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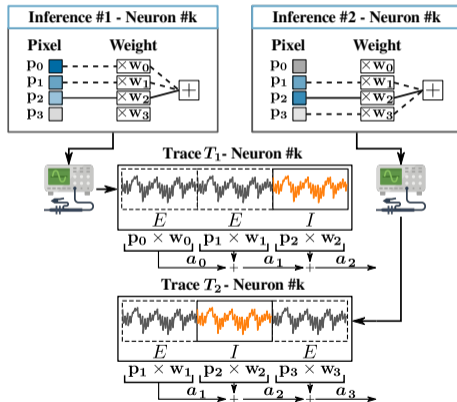


- What happens if we *prune* (skip) pixels?
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 - Non-important weight
 - Non-important Multiply-and-Accumulate (NIMAC)
- **Important pixel**: important accuracy reduction
 - Important weight
 - Important MAC (IMAC)

MACPruning

The Idea

At each inference, randomly skip (*prune*) non-important pixels.

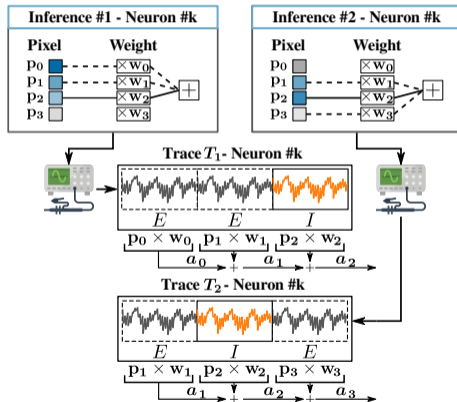


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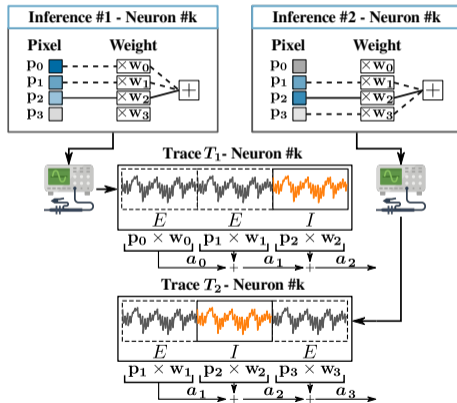
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Effects

- **Incorrect leakage predictions,**



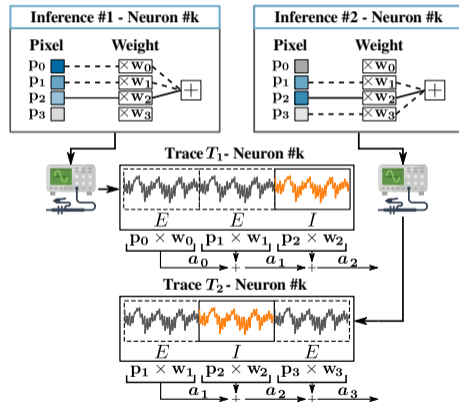
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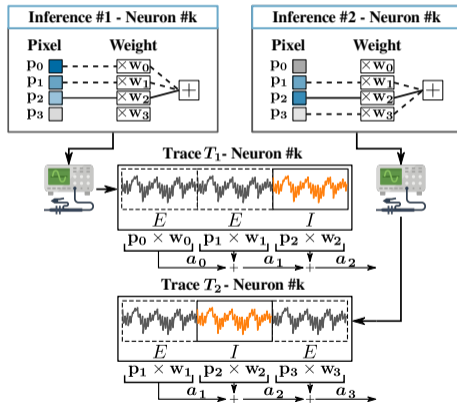
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Effects

- Incorrect leakage predictions,
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Security Implications

Successful weight recovery requires exponential number of traces.



MACPruning Implementation

Input : *inputs*: ($w \times h$)-array of pixels
weights: ($w \times h$)-array of weights
laPAM: ($w \times h$)-array of bits
randWords: ($w \times h$)-array of bits

Output: *acc*: accumulator

```
1 acc ← 0;  
2 for i from 0 to length(inputs) do  
   |  
   |  
   |  
10 end
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At each inference, randomly skip (*prune*) non-important pixels.

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- *laPAM*: pixel \mapsto importance,
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 - *i*-th pixel is important (Line 3–5) or,

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9   else
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11 end
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 - i -th pixel must be executed (Line 6–8),
- Control-flow-dependent check.

Control-flow Dependency and Side-channel Information

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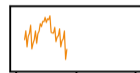
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Important



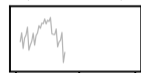
280 330 379

Non-Important
(Executed)



1276 1336 1395

Non-Important
(Skipped)



104 148 187

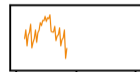
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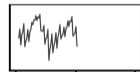
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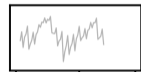
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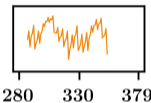
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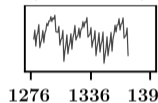
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```

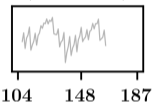
Important



Non-Important
(Executed)



Non-Important
(Skipped)



Control-flow Dependency and Side-channel Information

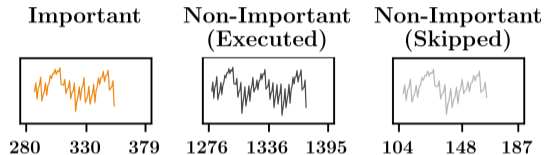
Input : *inputs*: ($w \times h$)-array of pixels
weights: ($w \times h$)-array of weights
laPAM: ($w \times h$)-array of bits
randWords: ($w \times h$)-array of bits

Output: *acc*: accumulator

```

1 acc ← 0;
2 for i from 0 to length(inputs) do
3   if laPAM[i] then
4     m ← inputs[i] · weights[i];
5     acc ← acc + m;
6   else if randWords[i] then
7     m ← inputs[i] · weights[i];
8     acc ← acc + m;
9   else
10    /* Skip MAC */
11 end
12 return acc

```



Side-Channel Information

Control-flow dependency hints each pixel' importance

Control-flow Dependency and Side-channel Information

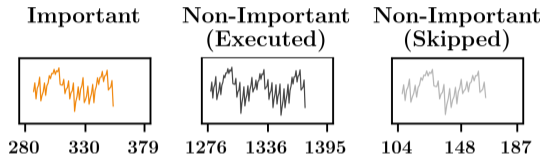
Input : *inputs*: ($w \times h$)-array of pixels
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```



Side-Channel Information

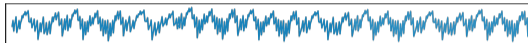
Control-flow dependency hints each pixel' importance

Question

What are the security implications?

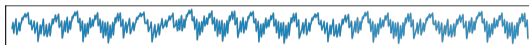
Control-flow Dependency and Security Implications

Given a side-channel trace ...

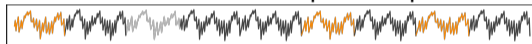


Control-flow Dependency and Security Implications

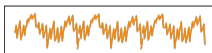
Given a side-channel trace ...



... we extract the important pixels.



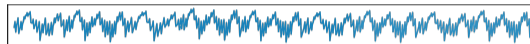
i_0 i_1 i_2 i_3 i_4 i_5 i_6 i_7 i_8



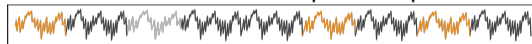
i_0 i_5 i_7

Control-flow Dependency and Security Implications

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i_0 i_1 i_2 i_3 i_4 i_5 i_6 i_7 i_8



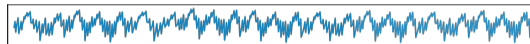
i_0 i_5 i_7



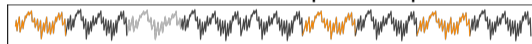
Circumvent trace desynchronisation

Control-flow Dependency and Security Implications

Given a side-channel trace ...



... we extract the important pixels.



i_0 i_1 i_2 i_3 i_4 i_5 i_6 i_7 i_8



i_0 i_5 i_7



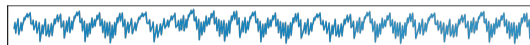
Circumvent trace desynchronisation

... we remove skipped pixels.

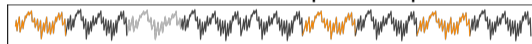


Control-flow Dependency and Security Implications

Given a side-channel trace ...



... we extract the important pixels.



i_0 i_1 i_2 i_3 i_4 i_5 i_6 i_7 i_8



i_0 i_5 i_7



Circumvent trace desynchronisation

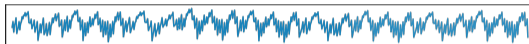
... we remove skipped pixels.



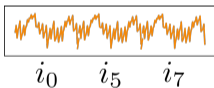
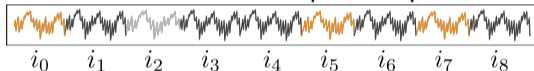
Compute correct leakage predictions

Control-flow Dependency and Security Implications

Given a side-channel trace ...



... we extract the important pixels.



Circumvent trace desynchronisation

... we remove skipped pixels.



Compute correct leakage predictions

Research Question

Can we circumvent the MACPruning countermeasure?

Table of Contents

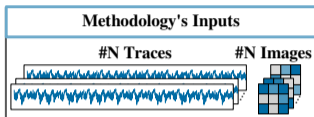
1 Background

2 Methodology

3 Results

4 Conclusions

A Preprocessing Methodology to Circumvent MACPruning



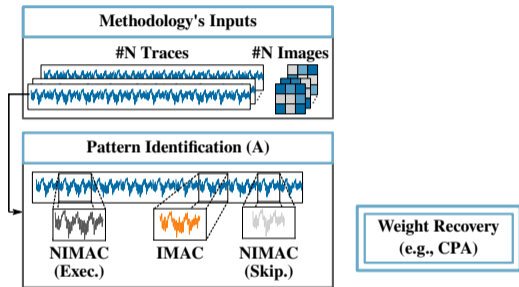
Weight Recovery
(e.g., CPA)

Goal
Recover the Important Weights.

Methodology

Figure: Preprocessing Methodology.

A Preprocessing Methodology to Circumvent MACPruning



Goal

Recover the Important Weights.

Methodology

- A. Identify Side-channel Patterns;

Figure: Preprocessing Methodology.

A Preprocessing Methodology to Circumvent MACPruning

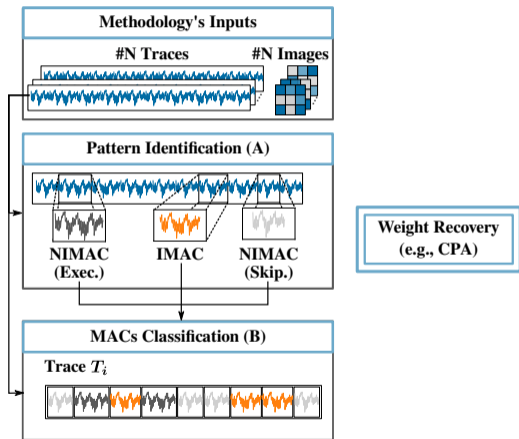


Figure: Preprocessing Methodology.

Goal

Recover the Important Weights.

Methodology

- A. Identify Side-channel Patterns;
- B. Identify MAC;/Inputs importance in traces;

A Preprocessing Methodology to Circumvent MACPruning

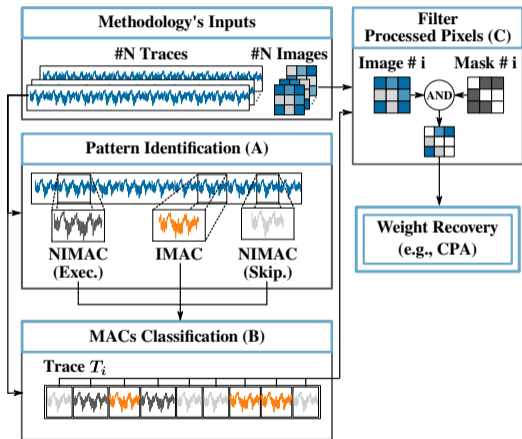


Figure: Preprocessing Methodology.

Goal

Recover the Important Weights.

Methodology

- Identify Side-channel Patterns;
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A Preprocessing Methodology to Circumvent MACPruning

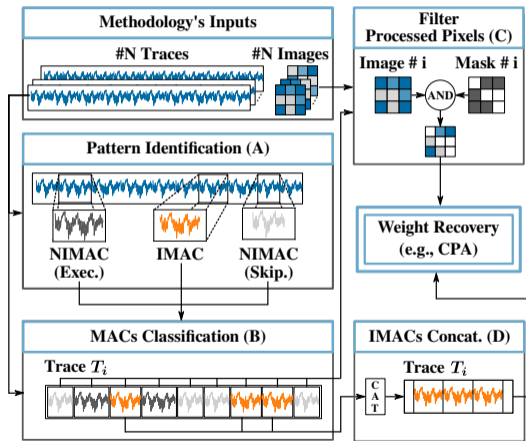


Figure: Preprocessing Methodology.

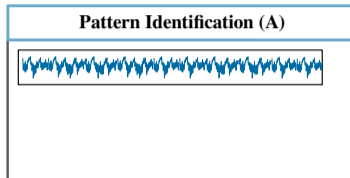
Goal

Recover the Important Weights.

Methodology

- Identify Side-channel Patterns;
- Identify MAC;/Inputs importance in traces;
- Remove skipped pixels from images;
- Preserve only the important MACs.

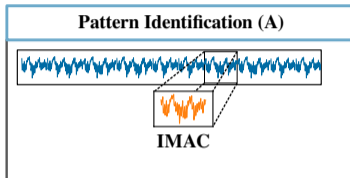
Step A – Pattern Identification



Pattern Identification

- Given one (or more) trace(s)...
- Identify a pattern for:

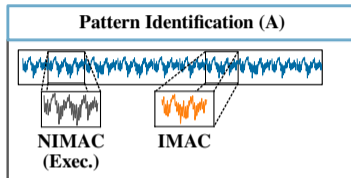
Step A – Pattern Identification



Pattern Identification

- Given one (or more) trace(s)...
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 - Important MACs (IMAC);

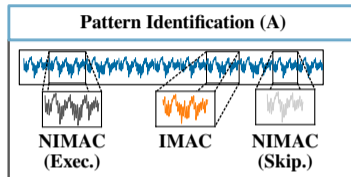
Step A – Pattern Identification



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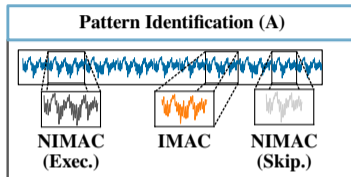
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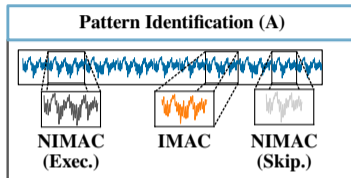
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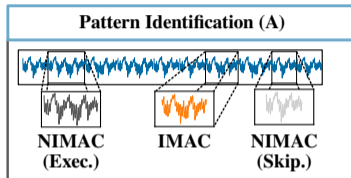


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Multiple patterns

Step A – Pattern Identification



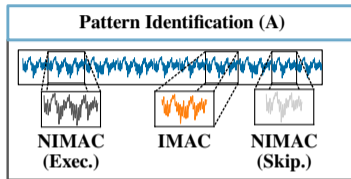
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Multiple patterns

- **Caveat:** one MAC may have multiple patterns

Step A – Pattern Identification



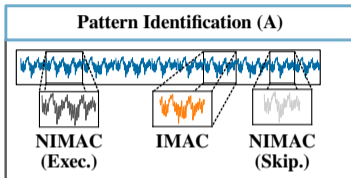
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 - Multiple instructions issue
 - Out-of-order architectures

Step A – Pattern Identification



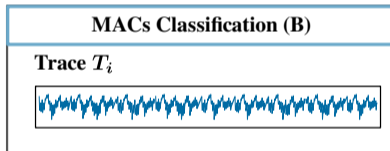
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Multiple patterns

- **Caveat:** one MAC may have multiple patterns
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 - Out-of-order architectures
- We assume to have 1 pattern per MAC (scalar architecture)

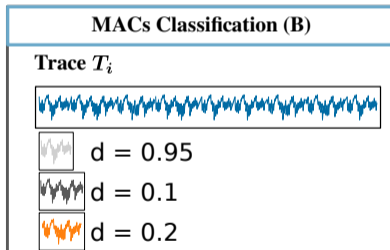
Step B – MACs Classification



Sliding-Window Pattern Matching

- 1 Define an acceptance threshold t ;
- 2 Sliding each pattern p on each trace sample s ;
- 3 For each pattern p , compute a distance score $d_p = d(p, T_i)$;
- 4 Assign to sample s pattern p if and only if:
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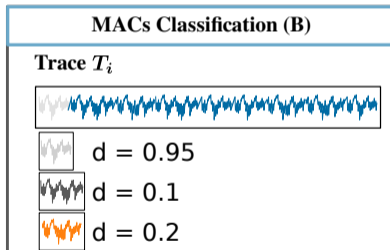
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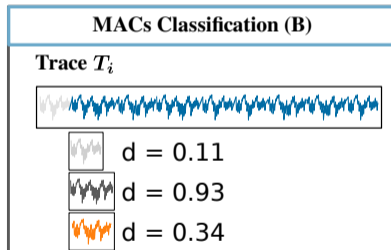
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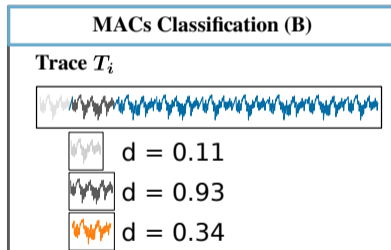
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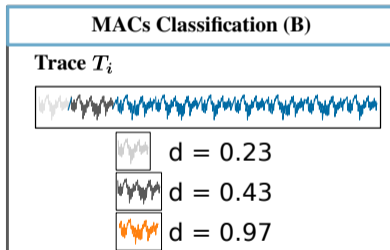
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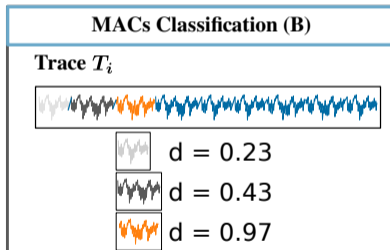
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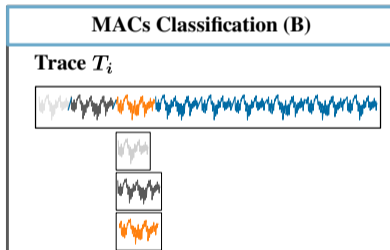
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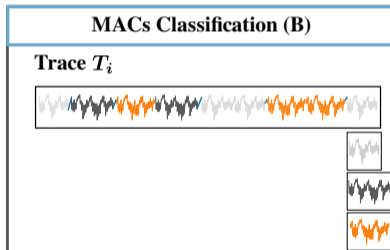
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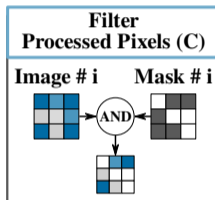
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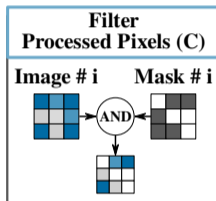
Step C – Filter Processed Pixels



Filter Processed Pixels

- Remove skipped pixels/MACs
- Allow for correct hypotheses computation

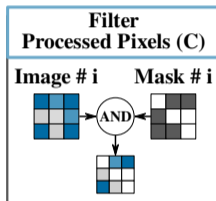
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Step C – Filter Processed Pixels

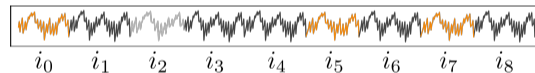
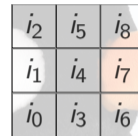
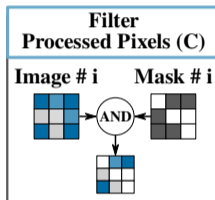


i_2	i_5	i_8
i_1	i_4	i_7
i_0	i_3	i_6

Filter Processed Pixels

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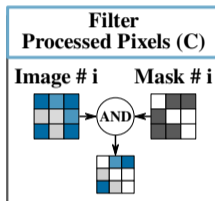
Step C – Filter Processed Pixels



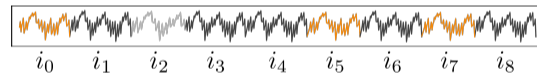
Filter Processed Pixels

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Step C – Filter Processed Pixels



i_2	i_5	i_8
i_1	i_4	i_7
i_0	i_3	i_6

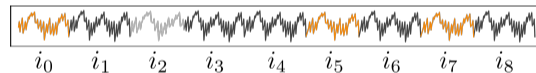
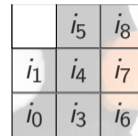
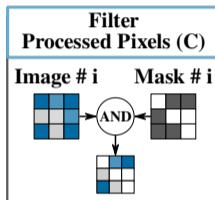


Filter Processed Pixels

- Remove skipped pixels/MACs
- Allow for correct hypotheses computation

To recover w_5 :

Step C – Filter Processed Pixels

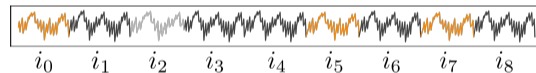
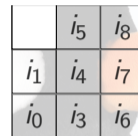
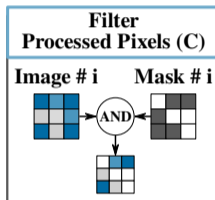


Filter Processed Pixels

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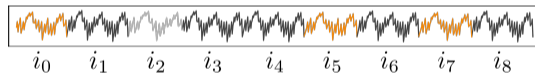
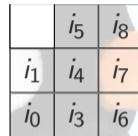
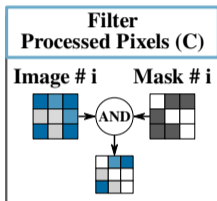
Filter Processed Pixels

- Remove skipped pixels/MACs
- Allow for correct hypotheses computation

To recover w_5 :

$$w_0 \cdot i_0 + i_1 \cdot w_1 + i_2 \cdot w_2 + i_3 \cdot w_3 + i_4 \cdot w_4 + i_5 \cdot w_5$$

Step C – Filter Processed Pixels



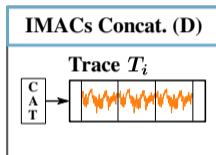
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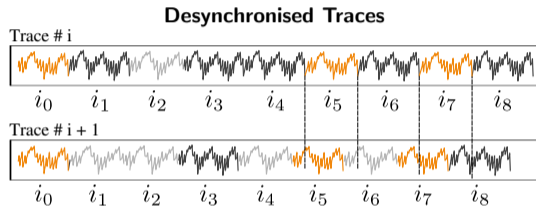
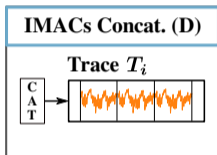
Step D – IMACs Concatenation



IMACs Concatenation

- Our goal: recover the Important MACs

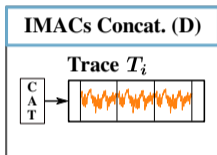
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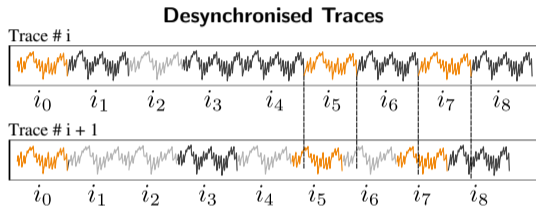
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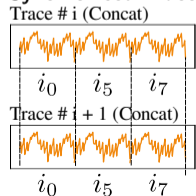


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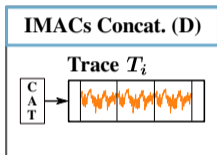
- Our goal: recover the Important MACs
- Preserve only the related patterns



Synchronised Traces



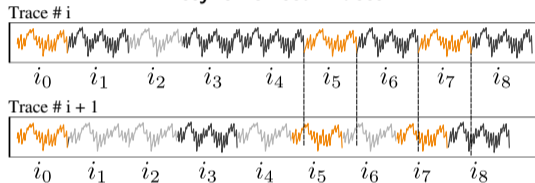
Step D – IMACs Concatenation



IMACs Concatenation

- Our goal: recover the Important MACs
- Preserve only the related patterns
- By-product: traces vertically aligned

Desynchronised Traces



Synchronised Traces

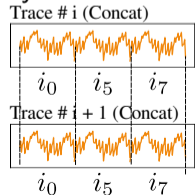


Table of Contents

1 Background

2 Methodology

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4 Conclusions

Experimental Setup

Deep Neural Network model

- Multi-Layer Preceptron
- 5 neurons in the first layer
- 32 pixels per neuron

Side-channel Setup

- ARM Cortex-M4 (Clk = 7.37MHz)
- Chipwhisperer Lite (Sampling rate = x4 Clk)

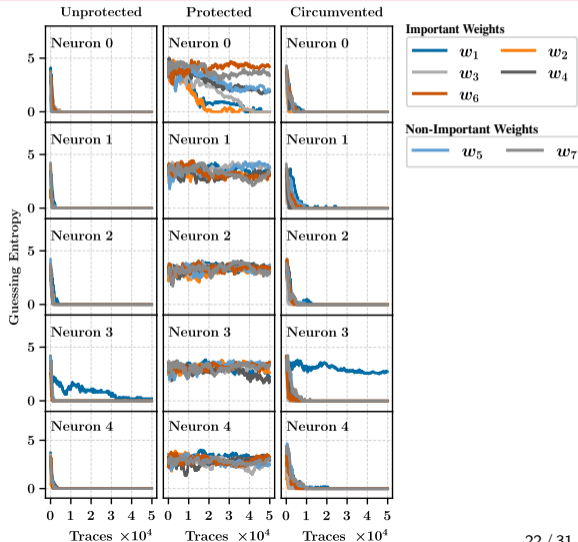
Experimental Procedure and Results

Experimental Procedure

- 1 Randomly and fix Important Pixels/MACs
- 2 Generate 5 sets of 50k random images
- 3 Recover 5 sets of 50k side-channel traces
- 4 Attack each set with Correlation Power Analysis
 - For each set, sort w_i 's values in descending order
- 5 Compute the *Guessing Entropy (GE)* of w_i :
 - Average position of the true weight value
 - If $GE(w_i) = 0 \rightarrow$ true weight scores the best

Results

- Recovered all the important weights



Summary Pt. 1

MACPruning:

Exponentially increases number of traces for successful weight extraction

Summary Pt. 1

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MACPruning Implementation:

Control-flow dependency (CFD) on pixel's importance

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Summary Pt. 1

Question:

What if we remove the control-flow dependency?

MACPruning: Fundamentally Broken?

Algorithm 4: Pseudo-code of a generic inference procedure enhanced with a control-flow-free MACPRUNING implementation.

Input : *inputs*: ($w \times h$)-array of pixels
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Output: *acc*: accumulator

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A Branchless MACPruning

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- **skip()**: simply return *acc*
- **exec()** and **skip** different by design

MACPruning: Fundamentally Broken?

Another Side-Channel Leakage

- `exec()` and `skip` different by design

MACPruning: Fundamentally Broken?

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- We identify important weights,
- Confidence bound on the number of traces.

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Cannot We Use (MAC)Pruning at all?

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Can Approximate Computing Underpin a Countermeasure?

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Approximate Computing is about *accuracy-performance* trade-off
Secure Approximate Computing might trade off accuracy for *security*

Thank You!

Questions?



Bibliography I

- [Din+25] Ruyi Ding et al. “MACPruning: Dynamic Operation Pruning to Mitigate Side-Channel DNN Model Extraction”. In: *2025 IEEE International Symposium on Hardware Oriented Security and Trust (HOST)*. 2025 IEEE International Symposium on Hardware Oriented Security and Trust (HOST). May 2025, pp. 12–22. DOI: 10.1109/HOST64725.2025.11050072. URL: <https://ieeexplore.ieee.org/document/11050072> (visited on 10/05/2025).

Bibliography II

- [Jap+25] Aditya Japa et al. “Security of Approximate Neural Networks against Power Side-channel Attacks”. In: *2025 62nd ACM/IEEE Design Automation Conference (DAC)*. 2025 62nd ACM/IEEE Design Automation Conference (DAC). June 2025, pp. 1–7. DOI: 10.1109/DAC63849.2025.11133333. URL: <https://ieeexplore.ieee.org/document/11133333/> (visited on 09/29/2025).
- [Leo+25] Vasileios Leon et al. “Approximate Computing Survey, Part I: Terminology and Software & Hardware Approximation Techniques”. In: *ACM Comput. Surv.* 57.7 (Mar. 5, 2025), 185:1–185:36. ISSN: 0360-0300. DOI: 10.1145/3716845. URL: <https://dl.acm.org/doi/10.1145/3716845> (visited on 04/22/2025).

Bibliography III

- [Zha+26] Lu Zhang et al. “ApproPower: Characterizing the Power Side-Channel Features of Approximate Multipliers with Symbolic Path Analysis”. In: *IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems* (2026), pp. 1–1. ISSN: 1937-4151. DOI: 10.1109/TCAD.2026.3656474. URL: <https://ieeexplore.ieee.org/abstract/document/11359193> (visited on 01/26/2026).

Microarchitectural Leakage

Unsuccessful Attacks?

- Not possible to recover the weight w_7
- Why?
- Solution: consider extended traces

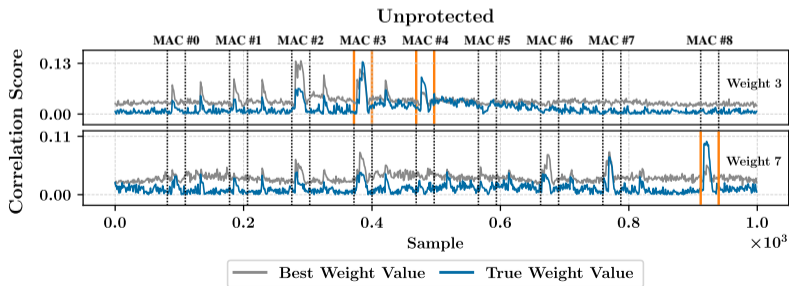
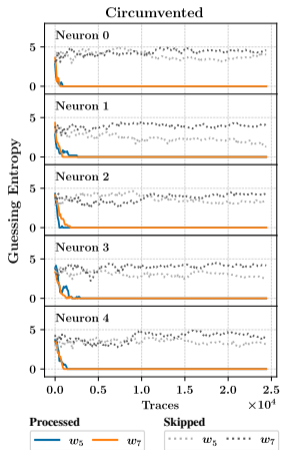


Figure: Microarchitectural Leakage.

Recovery of Non-Important Weights



Information carrying

- Partition traces in two sets:
 - *Processed*: non-important weight w_i processed
 - *Skipped*: non-important weight w_i skipped
- Compute GE on both sets
- Filtered traces carry information on non-important weights

Figure: GE Filtered Traces.