

Making Masking Security Proofs Concrete

Or How to Evaluate the Security of any Leaking Device



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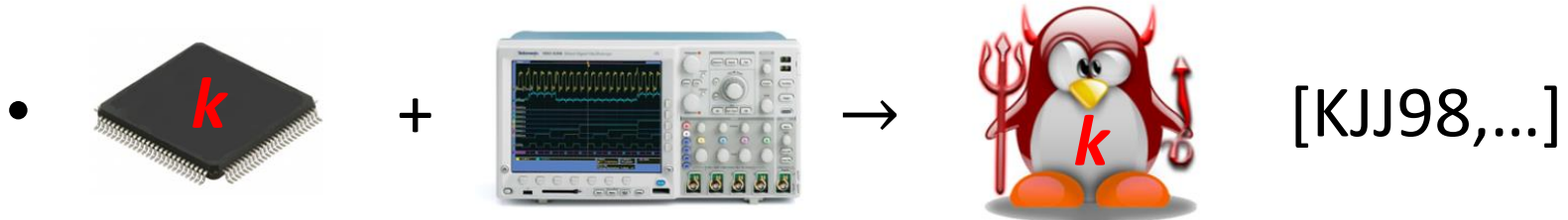
EUROCRYPT 2015, Sofia, Bulgaria

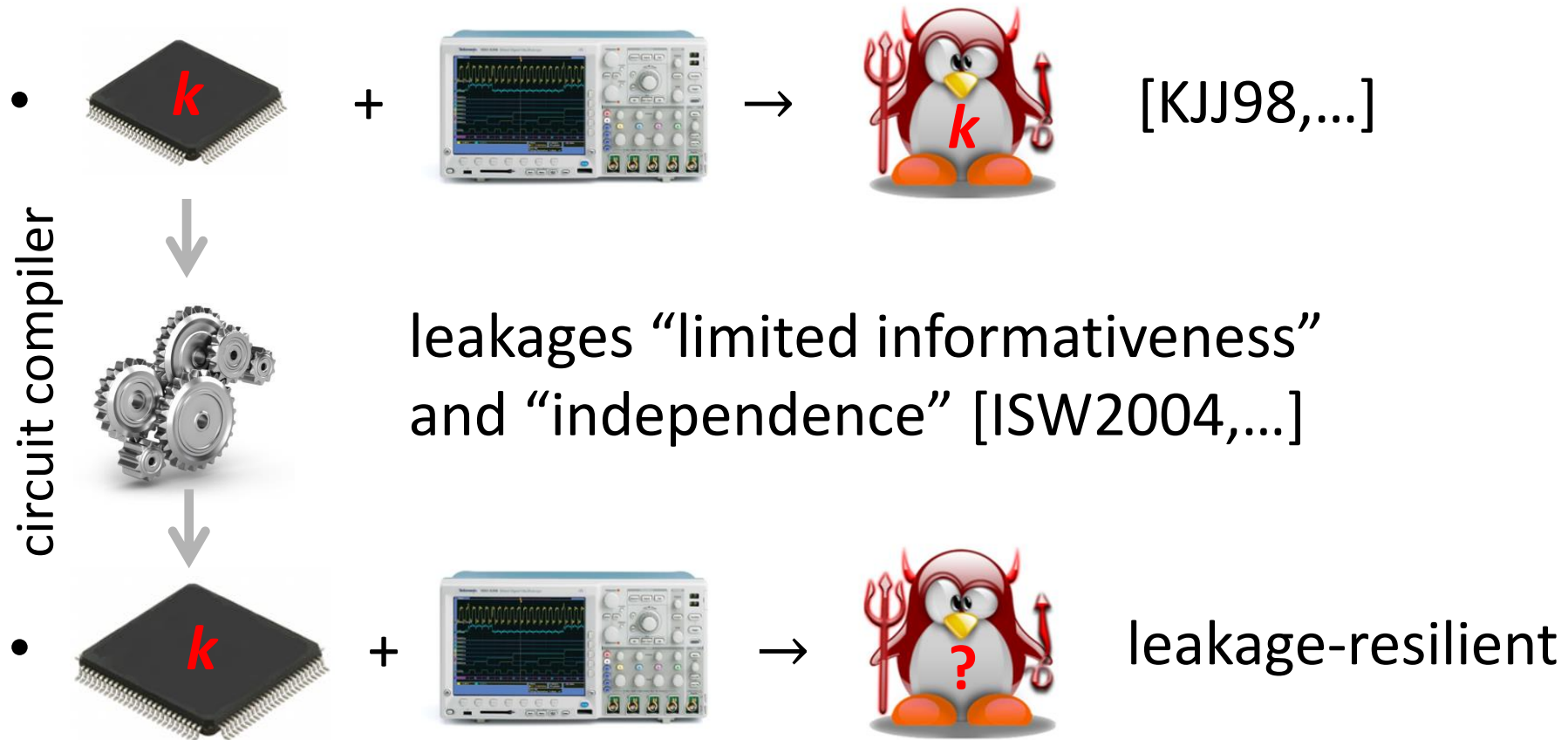
Outline: a collection of results...





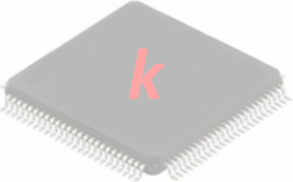


- Making masking proofs concrete
 - \approx connecting two recent results
- Evaluating masking bounds' tightness
- Analyzing assumptions
 - Sufficiently noisy leakages
 - Independent leakages
- Quantifying computational security
- Putting things together

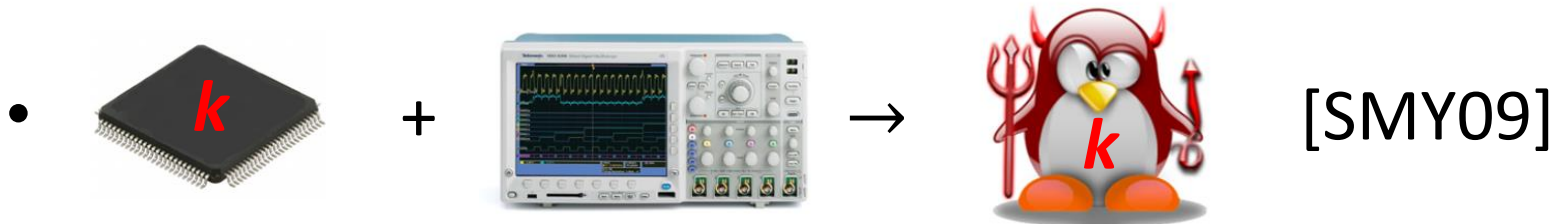
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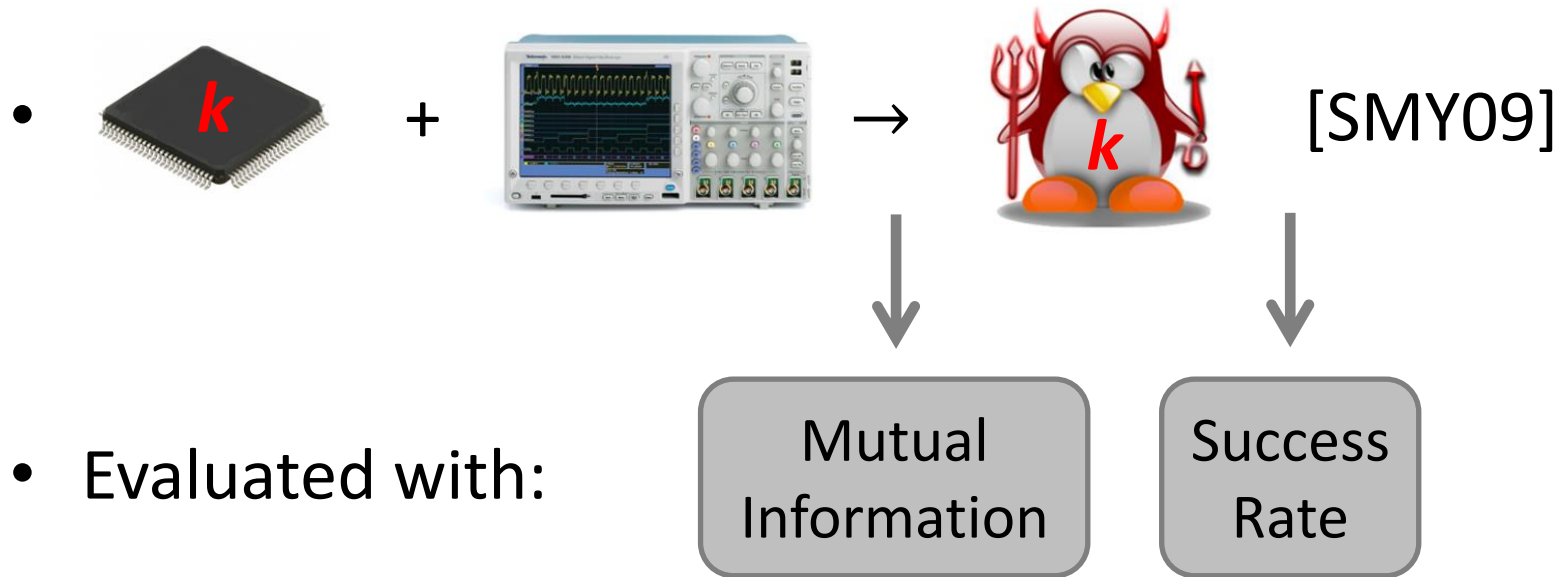
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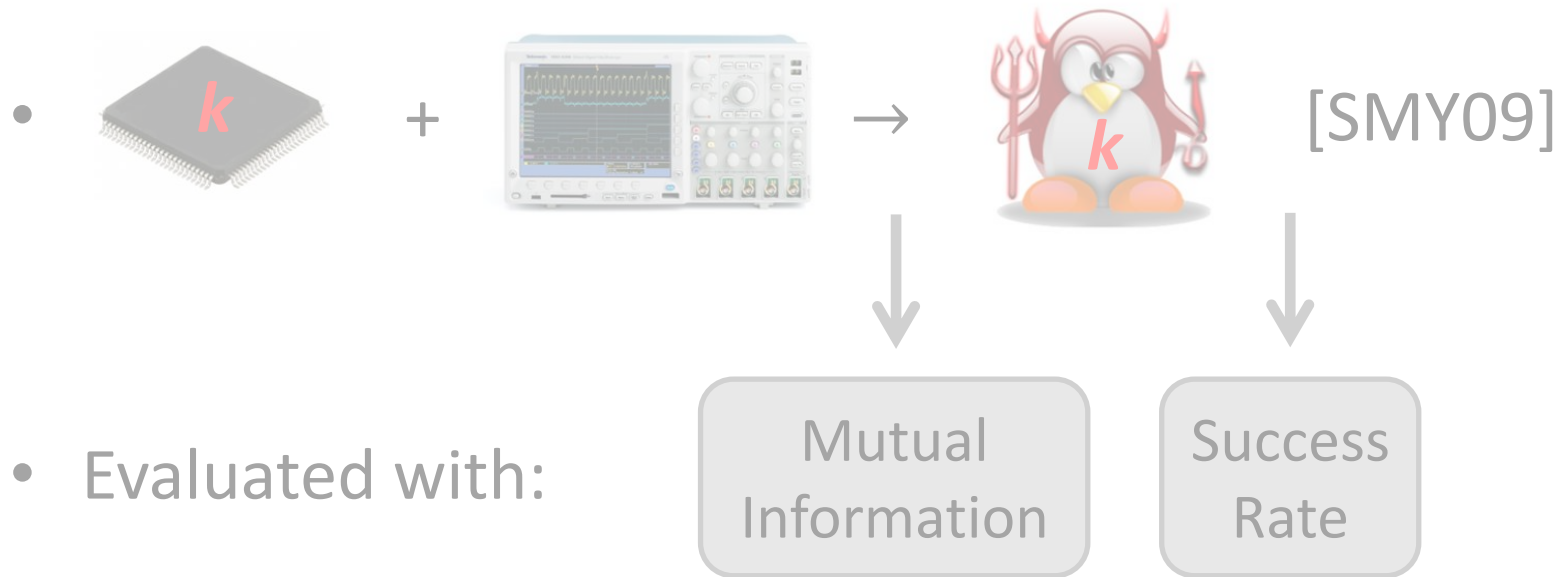




-  +  →  [KJJ98,...]
- circuit compiler
↓

↓
 +  →  leakage resilient
- [DDF2014]: reduction from conceptually simple probing model to more realistic noisy leakages model [PR13]

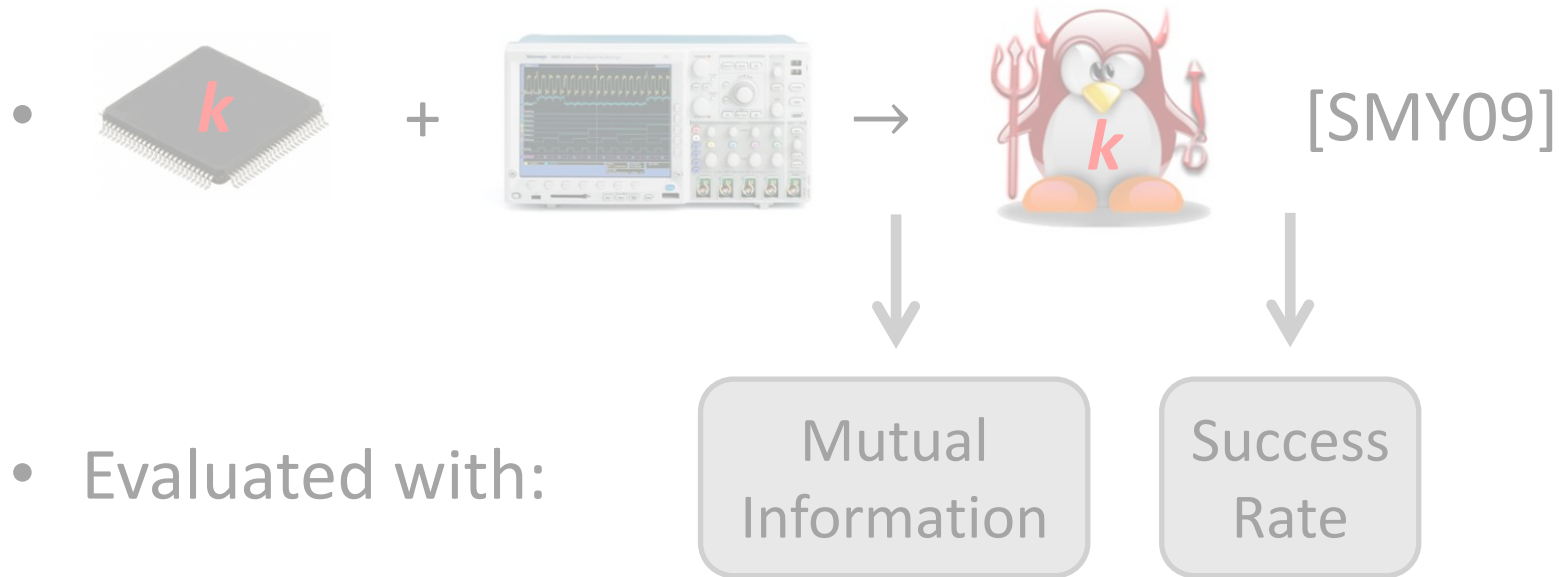






T1. Informative leakages => successful attacks ✓

T2. Link between mutual information and success rate ?



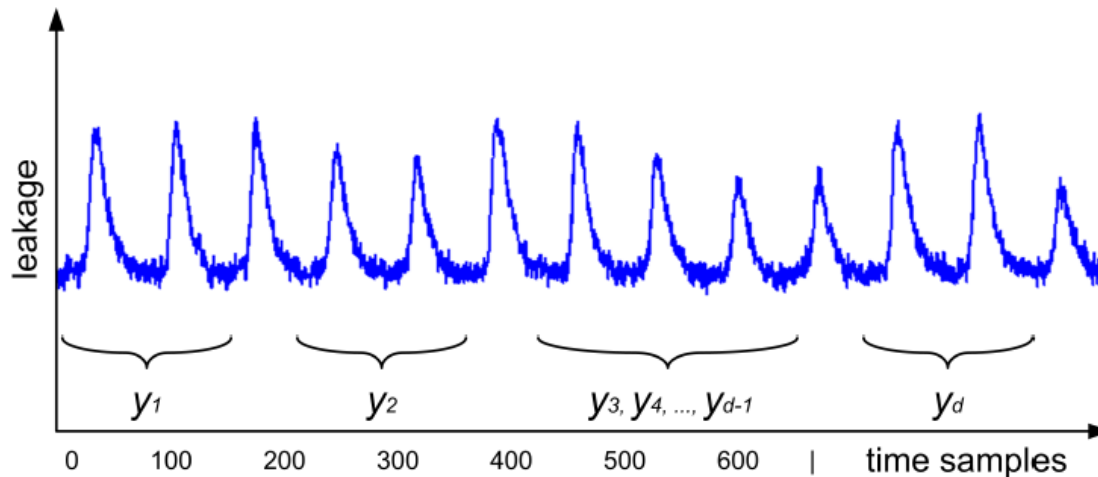
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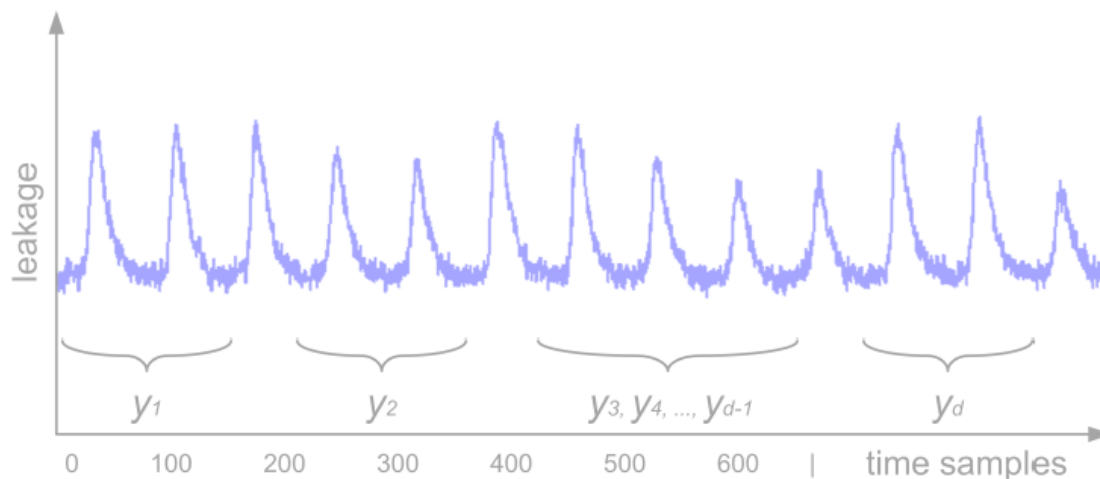
- Applied to many implementations/countermeasures

- Let $y = S(x \oplus k)$ be a leaking S-box computation
- Let $y = y_1 \oplus y_2 \oplus \cdots \oplus y_d$ be a sharing of y

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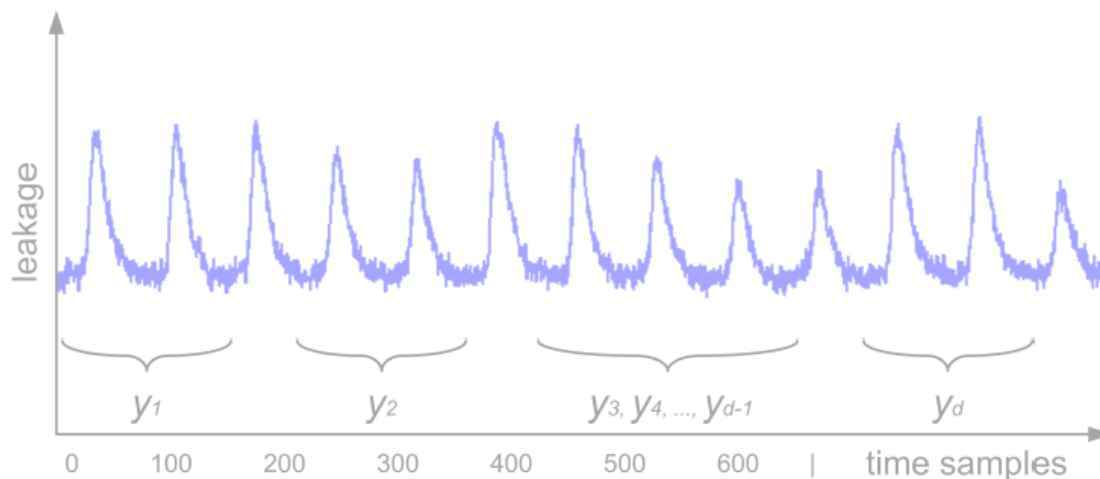


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- [DDF14] measures informativeness with $SD(Y_i; Y_i | L_{Y_i})$
- [SMY09] measures informativeness with $MI(Y_i | L_{Y_i})$

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- [DDF14] measures informativeness with $SD(Y_i; Y_i | L_{Y_i})$
- [SMY09] measures informativeness with $MI(Y_i | L_{Y_i})$
- [D12] showed that $2 \cdot SD(Y_i; Y_i | L_{Y_i})^2 \leq MI(Y_i | L_{Y_i})$

- Assume leakage variables $L_{Y_i} = L(Y_i, R_i)$ such that
 - $\text{MI}(Y_i | L_{Y_i}) \leq \frac{2}{|F|^2}$ (with $|F|$ the field size)
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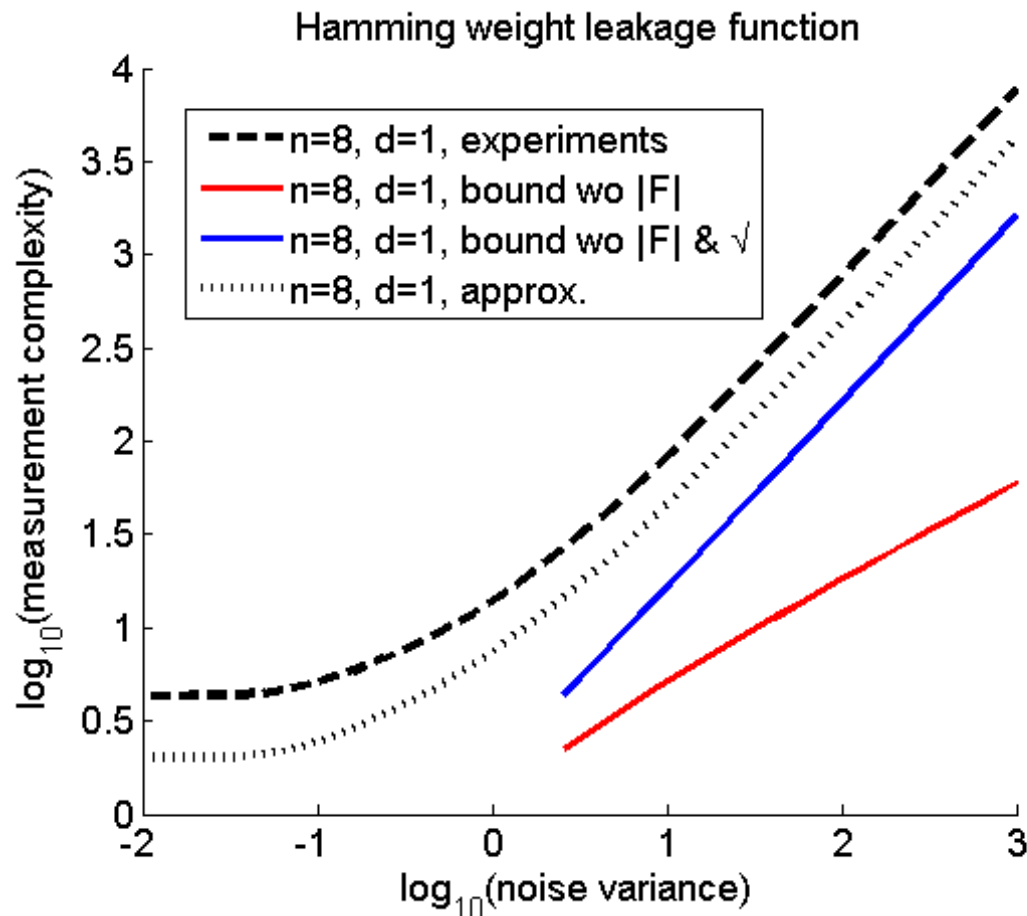
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- **Which (for $d=1$) proves T2 in [SMY09]**
- We provide a bound for complete circuits in the paper

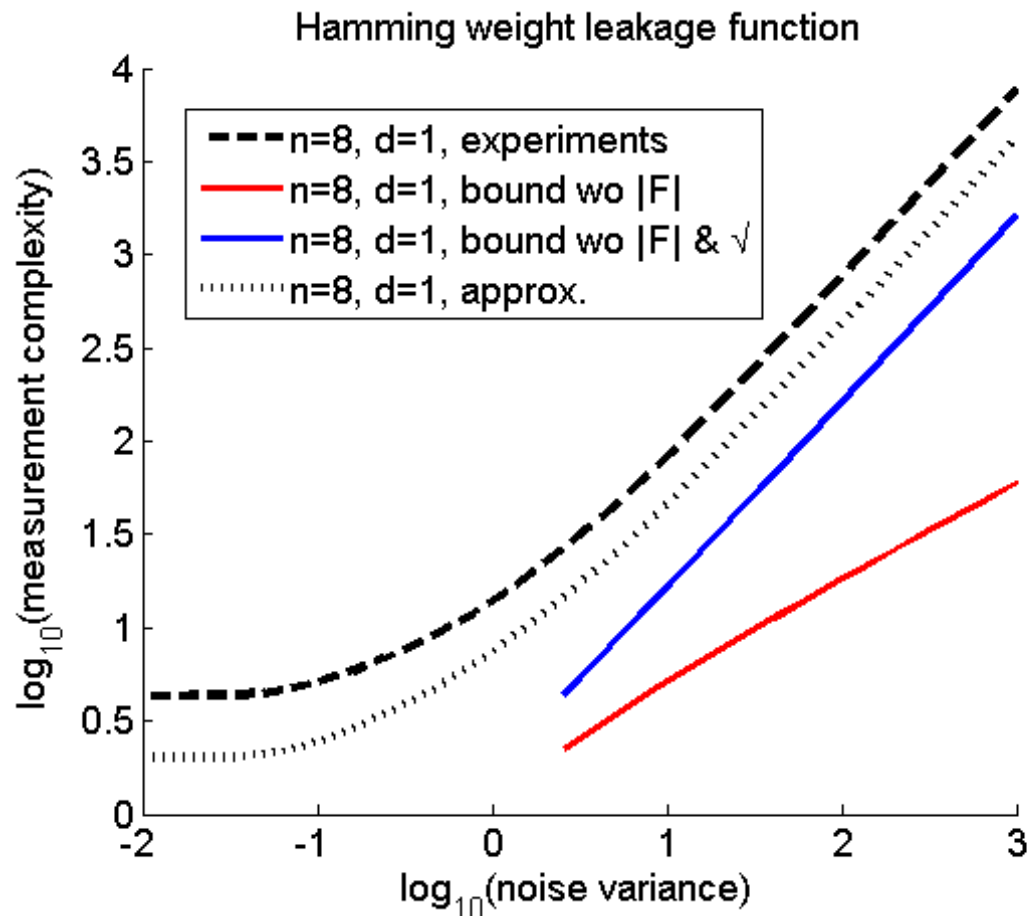
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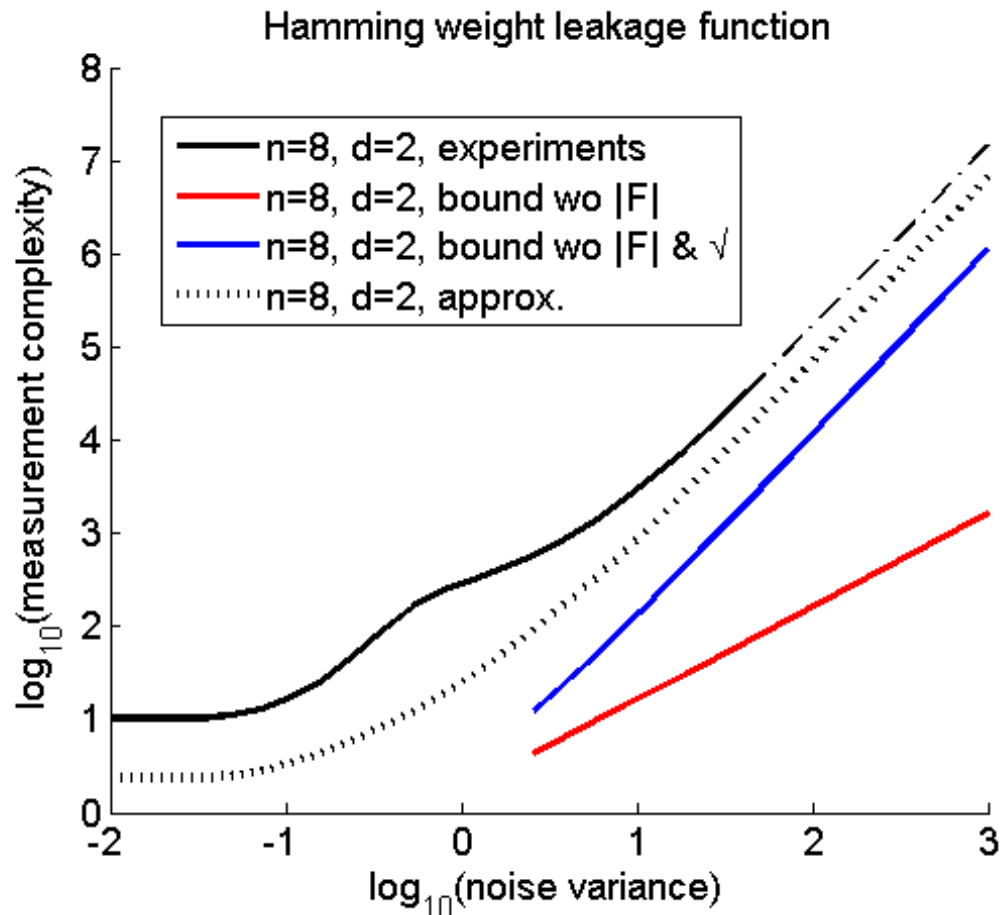


Unprotected case:

- $|F|$ factor \approx artifact (proven in [DFS15])
- $\sqrt{\cdot}$ loss \approx untight link between SD and MI
- [PR14] showed:

$$\text{MI} \leq \frac{|F|}{\ln(2)} \cdot \text{SD}$$

- $L_{Y_i} = \text{HW}(Y_i) + N_i$ with N_i Gaussian-distributed



Masked case:

- Similar results
- “Slope intuition”
(lowest moment of the distribution exploited)

- Under sufficiently noisy & independent leakages,

$$\text{SR} \leq 1 - \left(1 - \left(|F| \sqrt{\frac{\text{MI}(Y_i | L_{Y_i})}{2}} \right)^d \right)^m$$

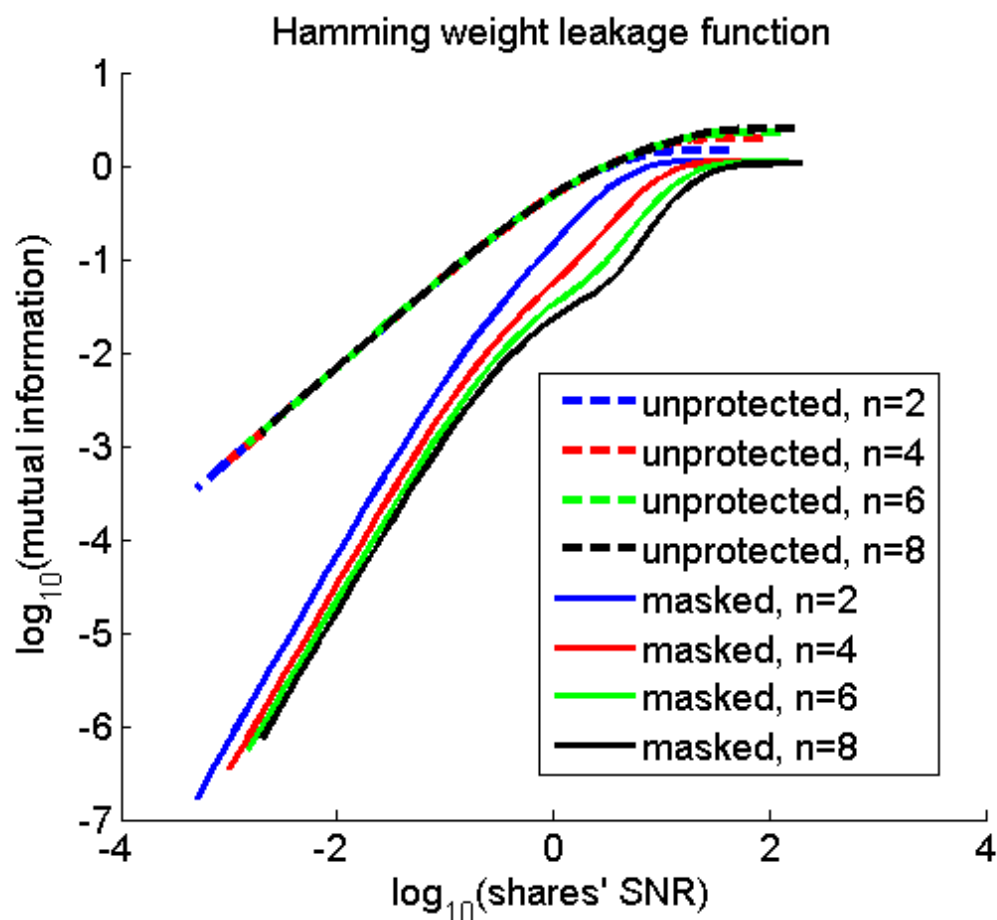
can be turned into: $\text{SR} \leq 1 - \left(1 - \text{MI}(Y_i | L_{Y_i})^d \right)^m$

- And this is also expected to hold for complete circuits

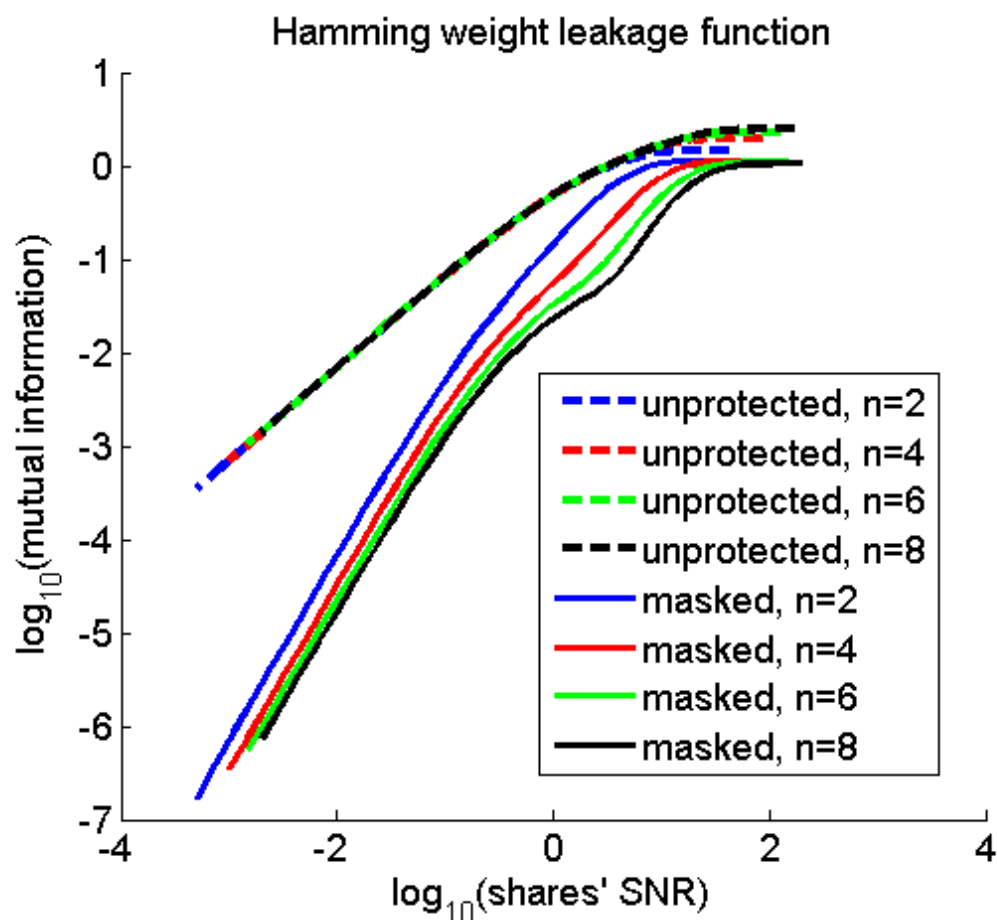
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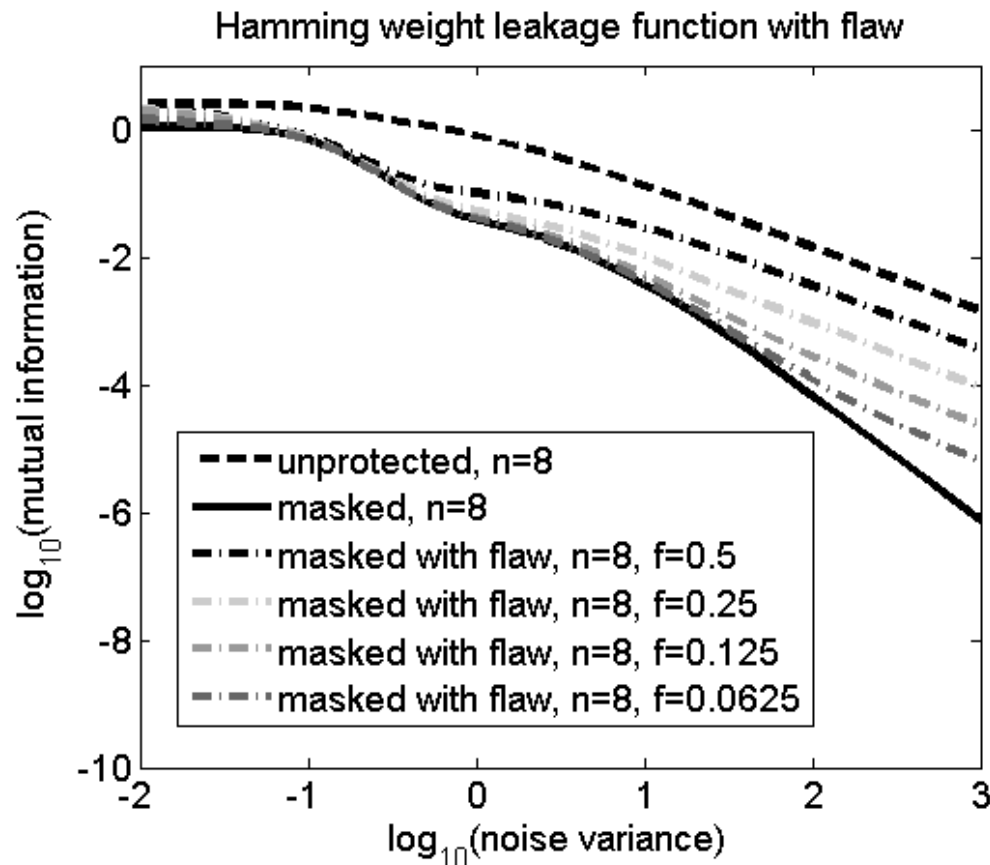


- “ Slope intuition”
- MI is small enough
when the signal
becomes significantly
smaller than the noise

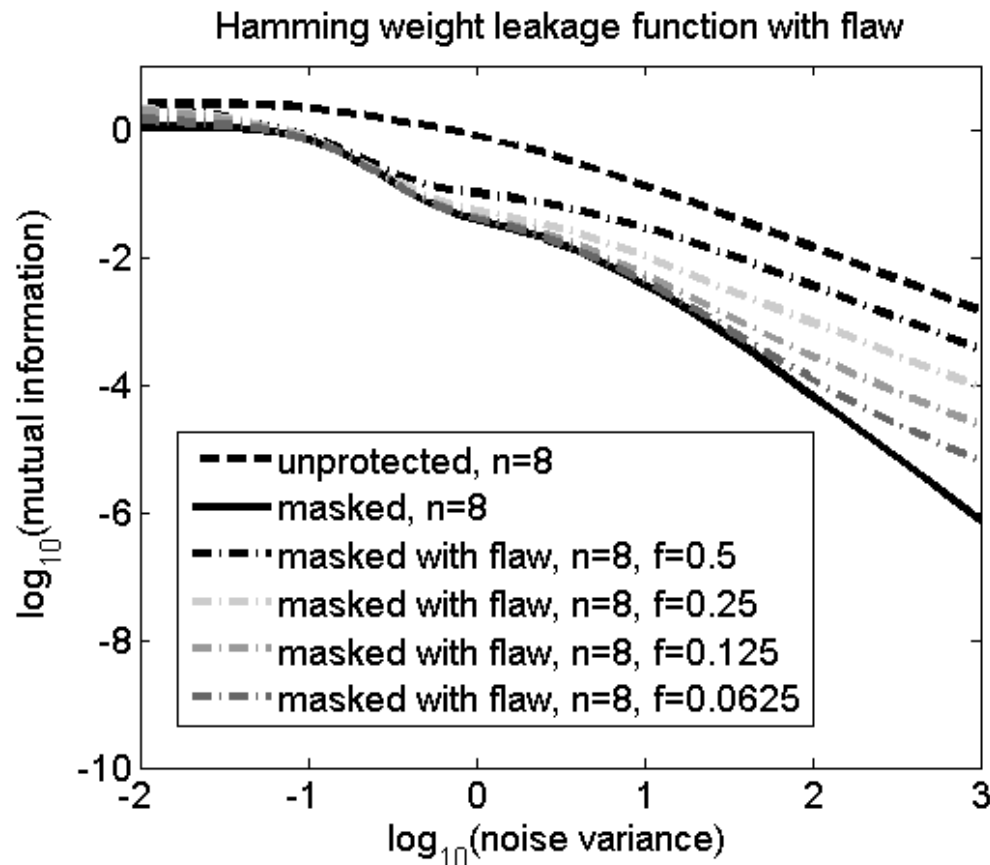
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- $L_{Y_2} = \text{HW}(Y_2) + f \cdot \text{HW}(Y_1 \oplus Y_2) + N_2$
- $f = 0$ means no flaw (independent leakages)

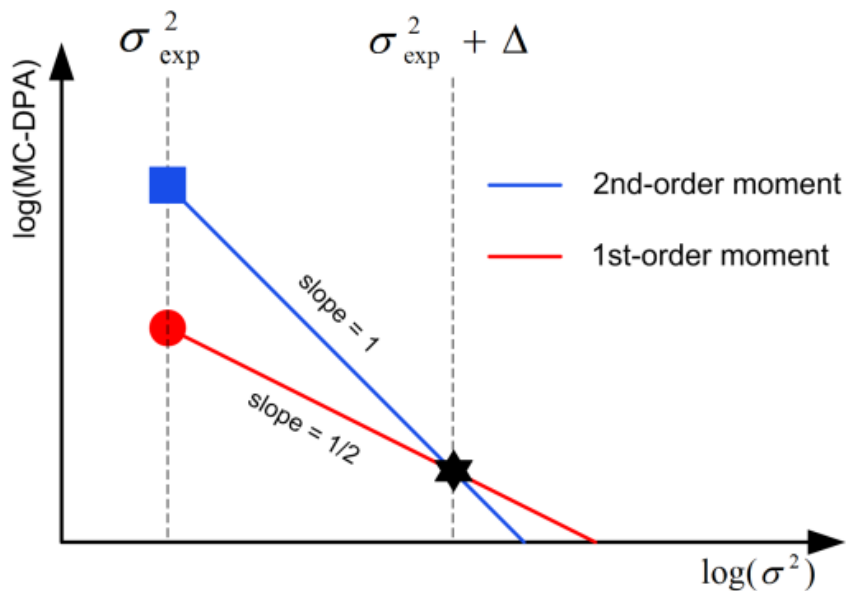


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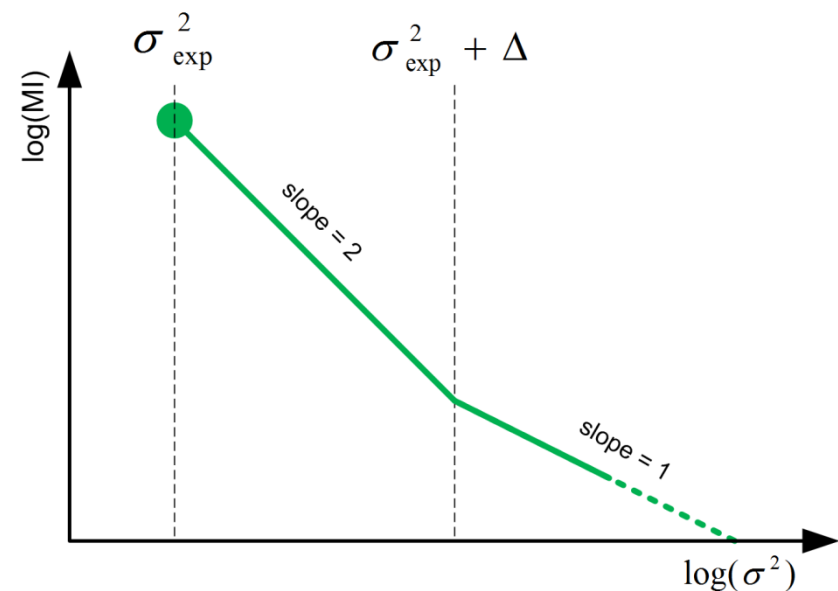
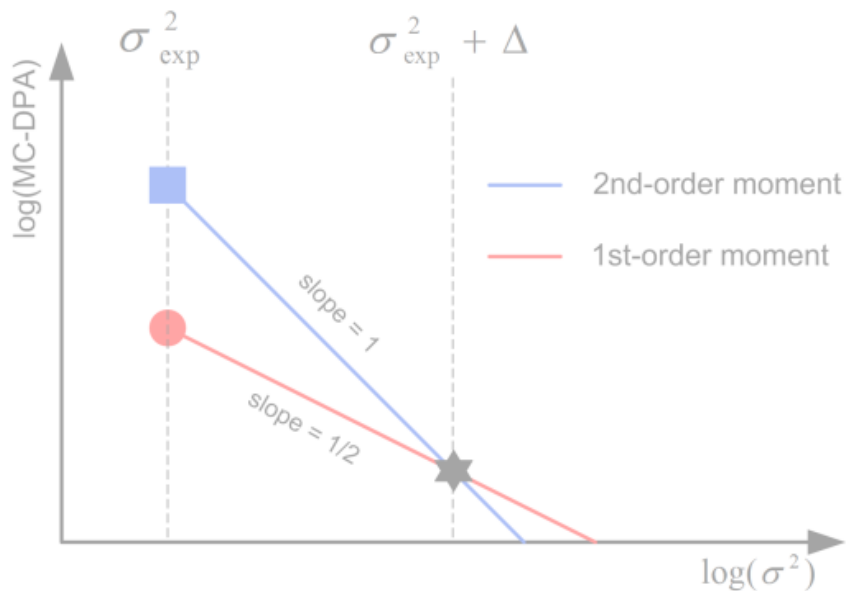
- Smaller flaws remain unnoticed until larger noises levels
- Masking with non-independent leakages improves security until certain noise levels
- “Slope intuition”

- For sufficiently noisy leakages (linear part of IT curves)
 - Evaluate the information “per moment”



- For sufficiently noisy leakages (linear part of IT curves)

1. Evaluate the information “per moment”



2. Extrapolate the impact on the MI

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- Attacker 1 (\approx probing model)
 - Learns nothing with $P=99/100$
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- Attacker 2 (\approx noisy leakage model)
 - Learns a set of 100 equally likely keys with $P=1$

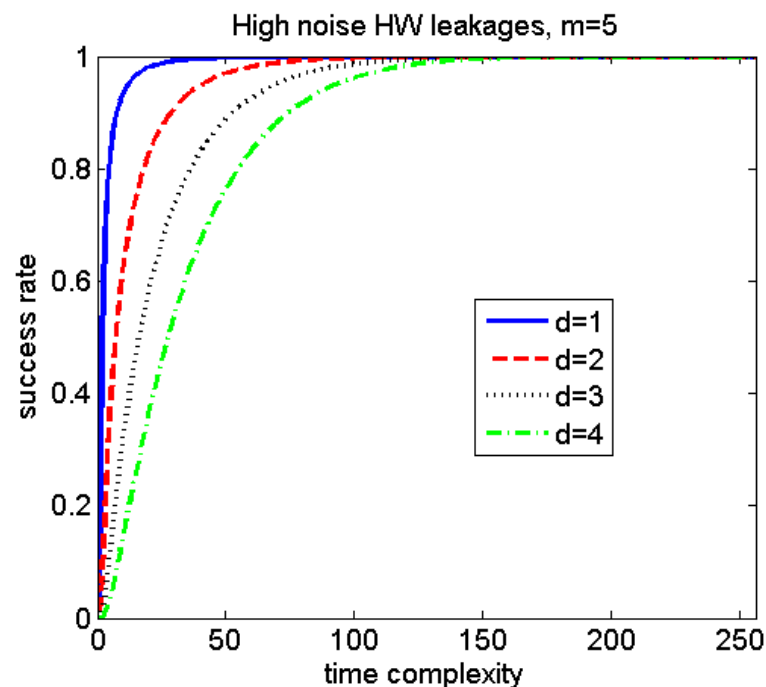
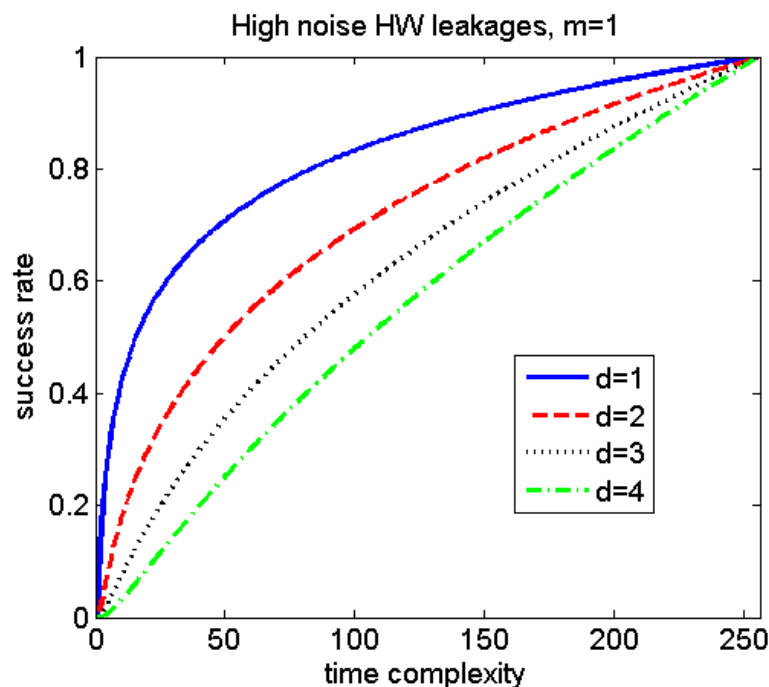
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=> Despite asymptotically equivalent, the probing model is better for proofs and the noisy leakage model is better for concrete evaluations considering computing power

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 - i.e. the MI on the c most likely candidates
- Our theorems directly bound the c 'th-order SR



- Similar to key rank estimation algorithms [VGS13]
- Problem can be written as:

$$\begin{array}{ll} \max_{c_1, \dots, c_{n_s}} & \sum_{i=1}^{n_s} \log(\text{SR}_i(m, c_i)) \\ \text{subject to} & \sum_{i=1}^{n_s} \log(c_i) \leq \log(\beta) \end{array}$$

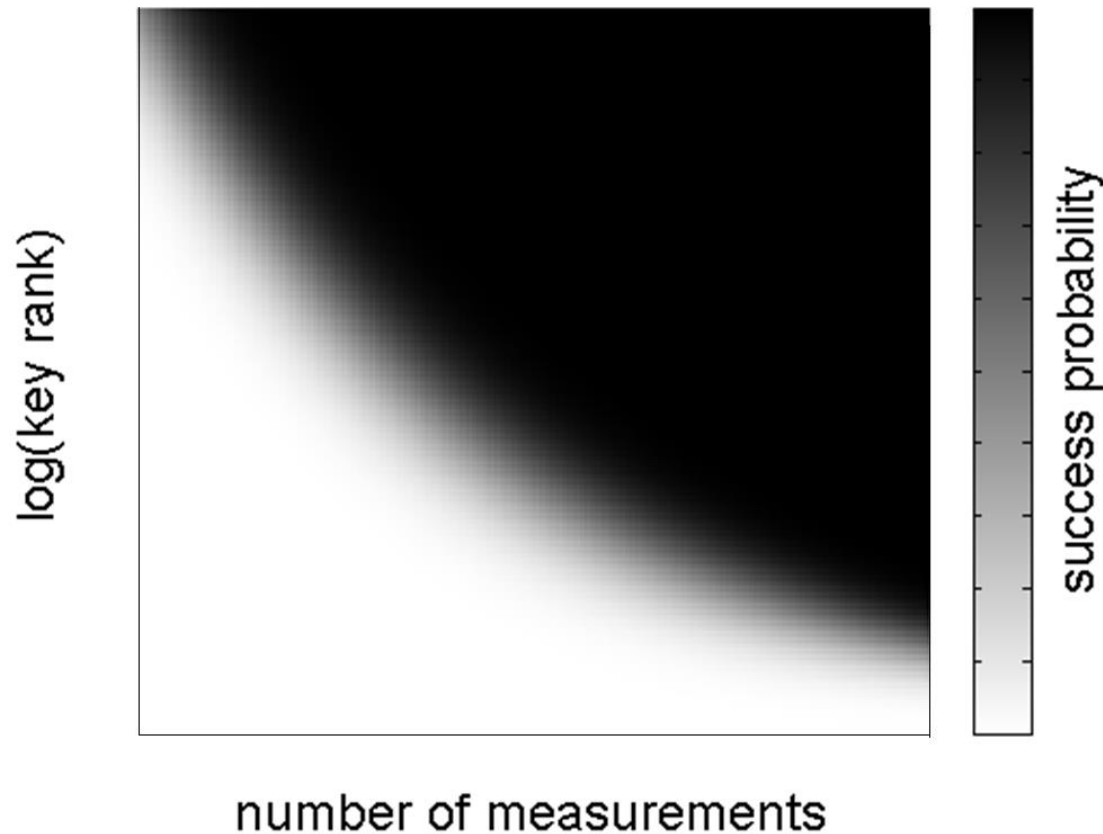
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- Non-linear knapsack / integer programming problem
- Many solutions in the literature (to be investigated)
- Cheap heuristics work well (done in the paper)
- (Also works “online” by replacing SRs by subkey prob.)

- Just repeat the previous procedure for various m 's



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- Existing approaches:



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DPA1(N^d traces), DPA2(N^d traces), ...



rank estimation #1
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...
rank estimation #Nr

1. exhaustive
evaluation



security graphs



Why is this helpful? (Masked device evaluation) 15

- Existing approaches:



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S-boxes SR



2. specific shortcut
[D+14,L+14]



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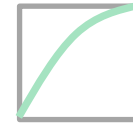
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security graphs  ,  , ...

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2. specific shortcut
[D+14,L+14]

worst-case bound
for security graphs

This paper:
generic shortcut



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 - If you care about the worst-case security level
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 - If you care about the worst-case security level
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- Proofs can be useful (to estimate concrete security)
- Open problems:
 - Investigating actual flaws (e.g. glitches)
 - Formalizing non-independent leakage
 - Maximum likelihood vs. weak maximum likelihood and nonlinear programming in rank estimation
 - Better (more secure and/or efficient) compilers

THANKS

<http://perso.uclouvain.be/fstandae/>