



Federal Office  
for Information Security



Royal Holloway  
University of London

# Optimal Recovery of Secret Keys from Weak Side Channel Traces

**Werner Schindler**

**BSI, Germany**

**Werner.Schindler@bsi.bund.de**

**Colin D. Walter**

**RHUL, UK**

**Colin.Walter@comodo.com**



# Outline

- The problem, background & history
- A typical randomised exponentiation algorithm
- The optimal decision strategy
  - General strategy and the Main Theorem
  - Example
- A metric to measure fitness of recoding guesses
- Results
- Conclusion



# Background

Several standard SW measures to counteract  
Side Channel Leakage from Exponentiation:

1. Blind the exponent by adding  
a random multiple of the group order.
2. Pick an algorithm where the pattern is  
independent of the secret exponent, e.g.
  - Square-and-always-multiply
  - Montgomery Powering Ladder
3. Use an algorithm where the pattern is randomised:
  - Liardet-Smart
  - Oswald-Aigner
  - Ha-Moon
  - Mist

We look at the strength of (3) with Ha-Moon for examples.



# Problems for an Attacker

- There is always a lot of noise in measurements.
- Averaging to determine correct key bits is essential.
- For randomised exponentiation algorithms, Square & Multiply operations cannot be aligned directly with key bits.
- Incorrect bit deductions will always occur.
- The locations of likely errors must be identified for a computationally feasible algorithm.



# Example: Ha-Moon

Recode the binary representation of key  $K$  from right to left:

- Add in the Carry of 0 or +1 to give new  $K$ .
- Choose digit 0 if  $K$  even.
- Randomly choose digit  $\pm 1$  if  $K$  odd.
- Set Carry to be 1 for digit  $-1$ , otherwise 0, & shift  $K$  down.

Exponentiation  $M^K$  in ECC:

- Repeatedly:
  - i) read next digit (from left to right)
  - ii) perform point double
  - iii) do point add if  $d = 1$  or point subtract if  $d = -1$ .

Traces may have different lengths: the  $i$ th operation is associated with different bits in different traces.



# Ha-Moon (II)

Here are some recodings of  $32n+13$ :

(**D** = double, **A** = add)

... 0 1 1 0 1	...	D D A D A D D A
... 0 1 1 1 -1	...	D D A D A D A D A
... 1 0 0 -1 -1	...	D A D D D A D A
... 1 0 -1 0 1	...	D A D D A D D A
... 1 0 -1 1 -1	...	D A D D A D A D A
... 1 -1 1 0 1	...	D A D A D A D D A

**Aim:** to recover  $K$  from leakage like the above.

The average operator yields almost no information:  
data from the top bit gets spread over several columns.



# History

Karlov & Wagner (CHES 2003)

Green, Noad & Smart (CHES 2005)

- Uses a Hidden Markov Model
- Applies Viterbi's algorithm to find the best fit key.
- Treats traces serially one by one
- Convergence is unlikely with weak leakage – it can't get started.

Walter (CHES 2008)

- Restructured to process traces in parallel, and bits serially
- Better convergence on weak leakage
- Lack of sound theoretical justification

Schindler (PKC 2005)

- Optimal decision strategy identifying most likely key
- Computationally infeasible in this context



# A Formal Approach (I)

Set of admissible keys:

- $\mathcal{K} \subseteq \mathbb{F}_2^*$

Set of all possible recoding sequences:

- $\mathcal{R} \subseteq \mathcal{D}^*$  where  $\mathcal{D} = \{\text{admissible recoding digits}\}$

Strategy (generic description):

- For each power trace  $pow_j$  ( $1 \leq j \leq N$ ) guess the individual recoding digits, yielding (disturbed, possibly invalid) noisy recoding sequences  $G_1, \dots, G_N \subseteq \mathcal{D}^*$ .
- Select the key  $K^*$  that fits  $G_1, \dots, G_N$  best.





# A Formal Approach (II)

Interpretation of the noisy recoding sequences  
as a two-step random experiment:

- $j^{\text{th}}$  randomised recoding sequence:  
$$\varphi: \mathcal{K} \times \mathcal{Y} \rightarrow \mathcal{R}, \quad \varphi(K, y_j) := R_j$$
where  $y_j$  is a random number. The target device contains a Finite Automaton to do this.
- $j^{\text{th}}$  noisy recoding guess:  
$$\psi: \mathcal{R} \times \mathcal{Z} \rightarrow \mathcal{D}^*, \quad \psi(R, z_j) := G_j$$
where  $z_j$  is a random number. The result of the adversary's inaccurate measurements.



# Main Theorem (I)

## Theorem 1(ii) (a special case):

### Assumptions:

- The unknown key  $K$  has been selected randomly according to some probability distribution  $\eta$
- Given recoding sequence guesses  $G_1, \dots, G_N$ .
- The adversary can detect whenever an operation of the recoded sequence  $R$  is carried out and guesses the types of these operations independently.

Notation:  $p(g|r) :=$

Prob(guesses  $op^n$  type is  $g$  given the true  $op^n$  type is  $r$ )

# Main Theorem (II)

## Theorem 1(ii) (special case, ctd'):

The optimal decision strategy selects a key  $K^* \in \mathcal{K}$  that maximises the term

$$\sum_{j=1}^N \log\left( \sum_{\substack{R \in R(K): \\ \text{len}(R) = \text{len}(G_j)}} \prod_{i=0}^{\text{len}(G_j)-1} p(g_{j,i} | r_i) \right)$$

assuming keys and recodings are distributed uniformly.

Note: Theorem 1(i) in the paper treats the most general case.



# Traces

The side channel gives a sequence of probabilities that the underlying operations correspond to particular digits.

- So we define a trace by  $T = (t_i)_{\{0 \leq i < \text{len}(G)\}}$   
with  $t_i =$  probability distribution on  $\mathcal{D}$   
(depending on the power trace)
- Thm 2 (a corollary of Thm 1 for traces) enables us to replace  $p(g_j|r)$  by  $t_i$  and so avoid guessing recodings  $G$ .



# Example (I)

## Application of Theorem 1(ii):

### Ha-Moon recoding with artificially small parameters:

- $\mathcal{K} = \{0, 1\}^n \setminus \{(0, \dots, 0)\}$ ,  $\mathcal{D} = \{‘S’, ‘M’, ‘\bar{M}’\}$

### Stochastic simulations

- Select  $K$  randomly
- Generate  $N$  recoding sequences  $R_1, \dots, R_N$
- Generate  $N$  noisy recoding sequences  $G_1, \dots, G_N$  by flipping recoding digits randomly
- More precisely:  
 $p(‘M’ | ‘S’) = 0.2$ ,  $p(‘\bar{M}’ | ‘S’) = 0.1$   
 $p(‘\bar{M}’ | ‘M’) = 0.2$ ,  $p(‘S’ | ‘M’) = 0.1$   
 $p(‘M’ | ‘\bar{M}’) = 0.2$ ,  $p(‘S’ | ‘\bar{M}’) = 0.1$



# Example (II)

## Application of Theorem 1(ii):

### Ha-Moon recoding with artificially small parameters:

- 100 stochastic simulations per table row

The correct key was ranked

Key length	# traces	1st	2nd	3rd - 9th	10th - 99th	100-999	>1000
15	10	84	5	9	1	1	0
20	10	57	20	20	2	1	0



## Example (III)

### Application of Theorem 1(ii): Ha-Moon recoding with artificially small parameters:

- These numerical results are remarkable since each recoding digit is correctly recognised despite probability of only 70% for individual operations!

However,

- unlike in many other side-channel attacks the optimal decision strategy cannot be applied to small portions of the key.
- Hence the application of Theorem 1 is infeasible for real-world key parameters.
- Starting from the optimal decision strategy a computationally feasible approximator is derived.

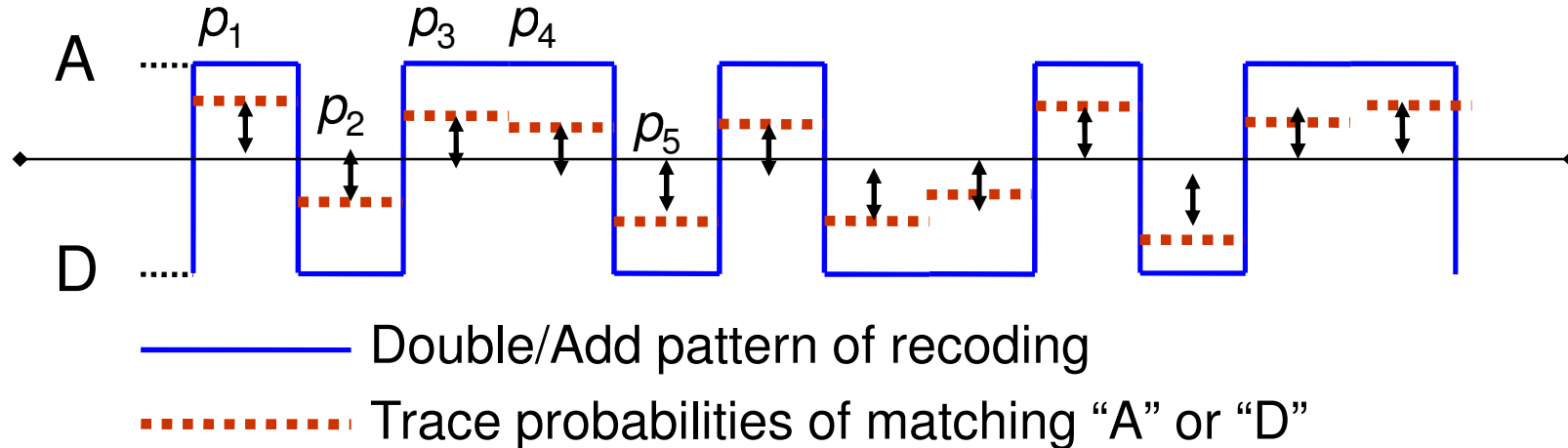


# The Metric (I)

Replace CHES08 distance between a trace  $t$  and recoding  $r$

$$\mu(t,r) = \sum_i (1-p_i) \quad \text{by "credibility"} \quad \mu(t,r) = \prod_i p_i$$

where  $p_i$  is probability that the  $i^{\text{th}}$  operation in  $t$  is the same as the  $i^{\text{th}}$  operation for  $r$ . (Hamming dist. vs Prob<sup>y</sup>.)



$$\mu = p_1 \times p_2 \times p_3 \times p_4 \times \dots$$



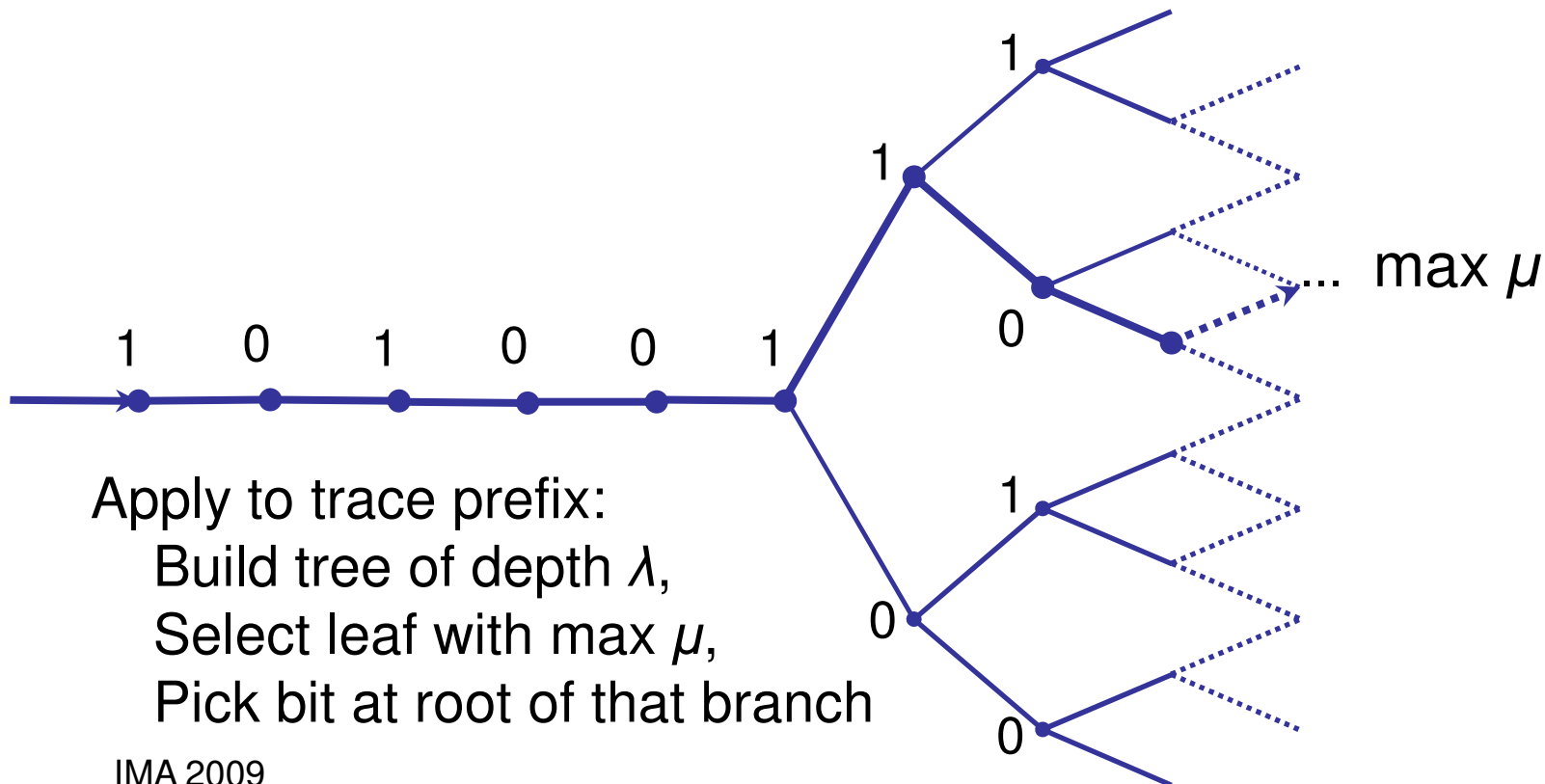


## The Metric (II)

- Define the **credibility** of key choice  $K$  for a trace  $t$  by

$$\mu(t, K) = \sum_r \{ \mu(t, r) \mid r \text{ is a recoding of } K \}$$

This selects the best match recoding of  $K$ .



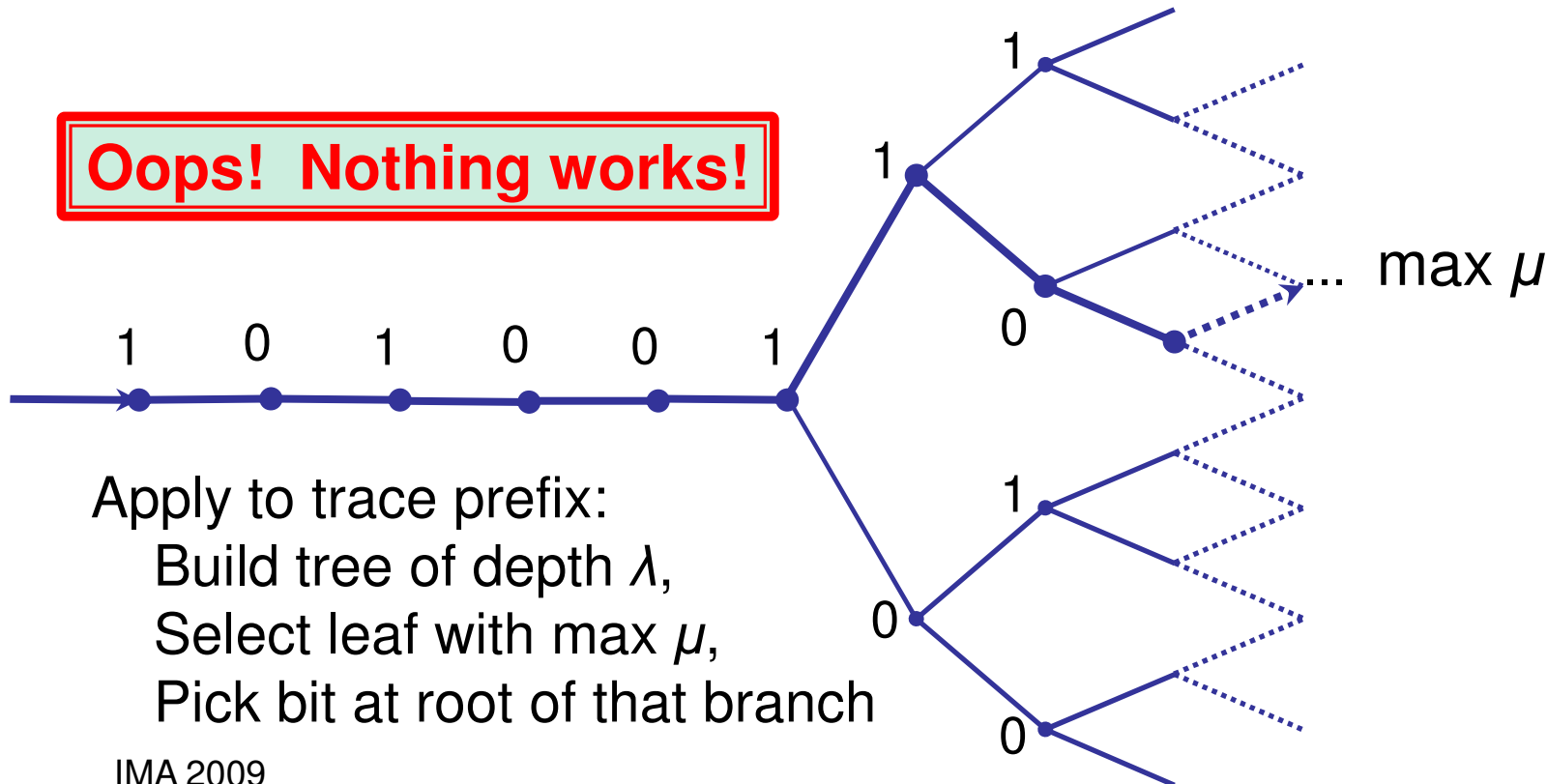
# The Metric (II)

- Define the **credibility** of key choice  $K$  for a trace  $t$  by

$$\mu(t, K) = \sum_r \{ \mu(t, r) \mid r \text{ is a recoding of } K \}$$

This selects the best match recoding of  $K$ .

**Oops! Nothing works!**



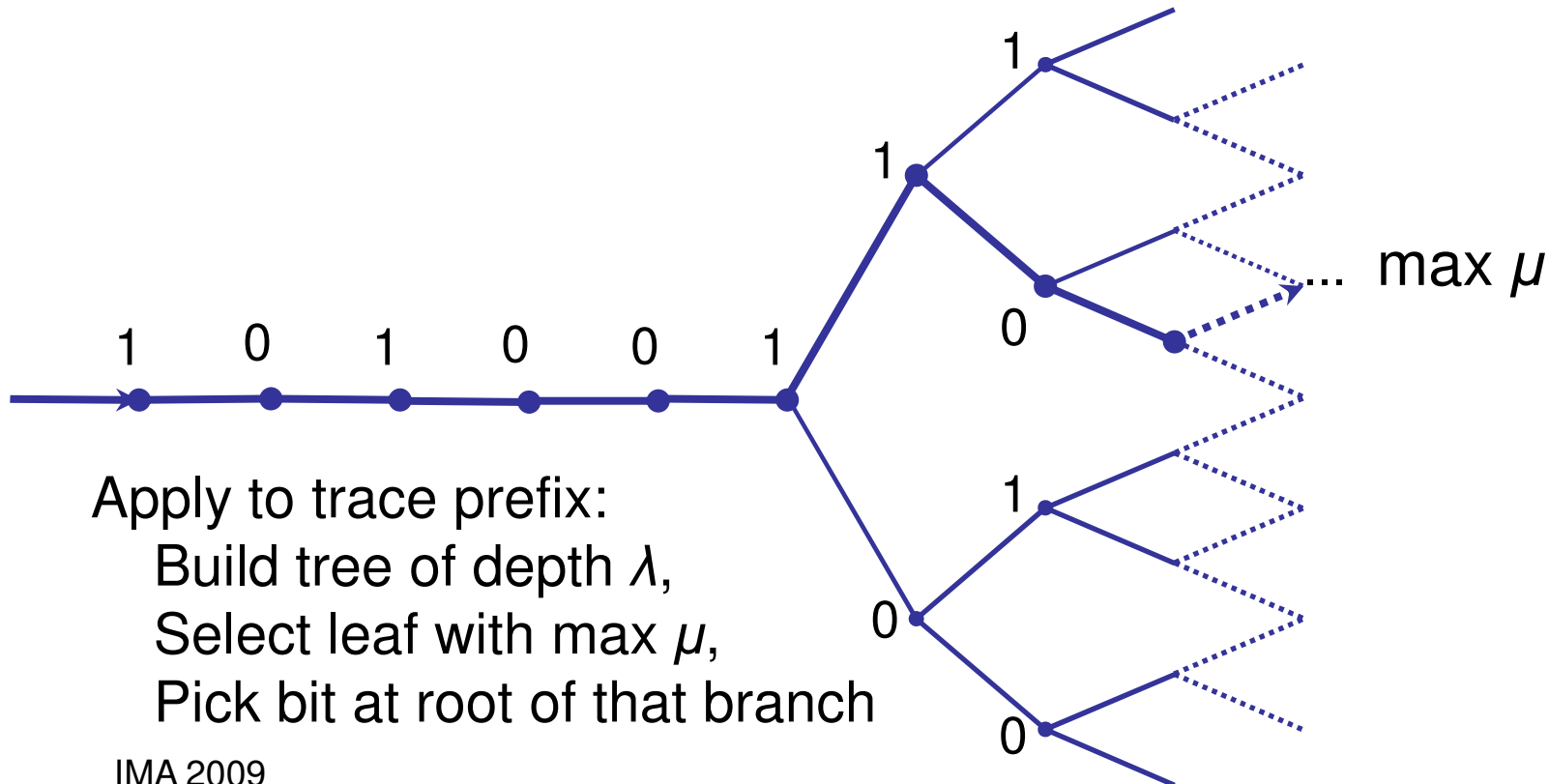


# The Metric (III)

- Modify the credibility definition, replacing “sum” by “max”:

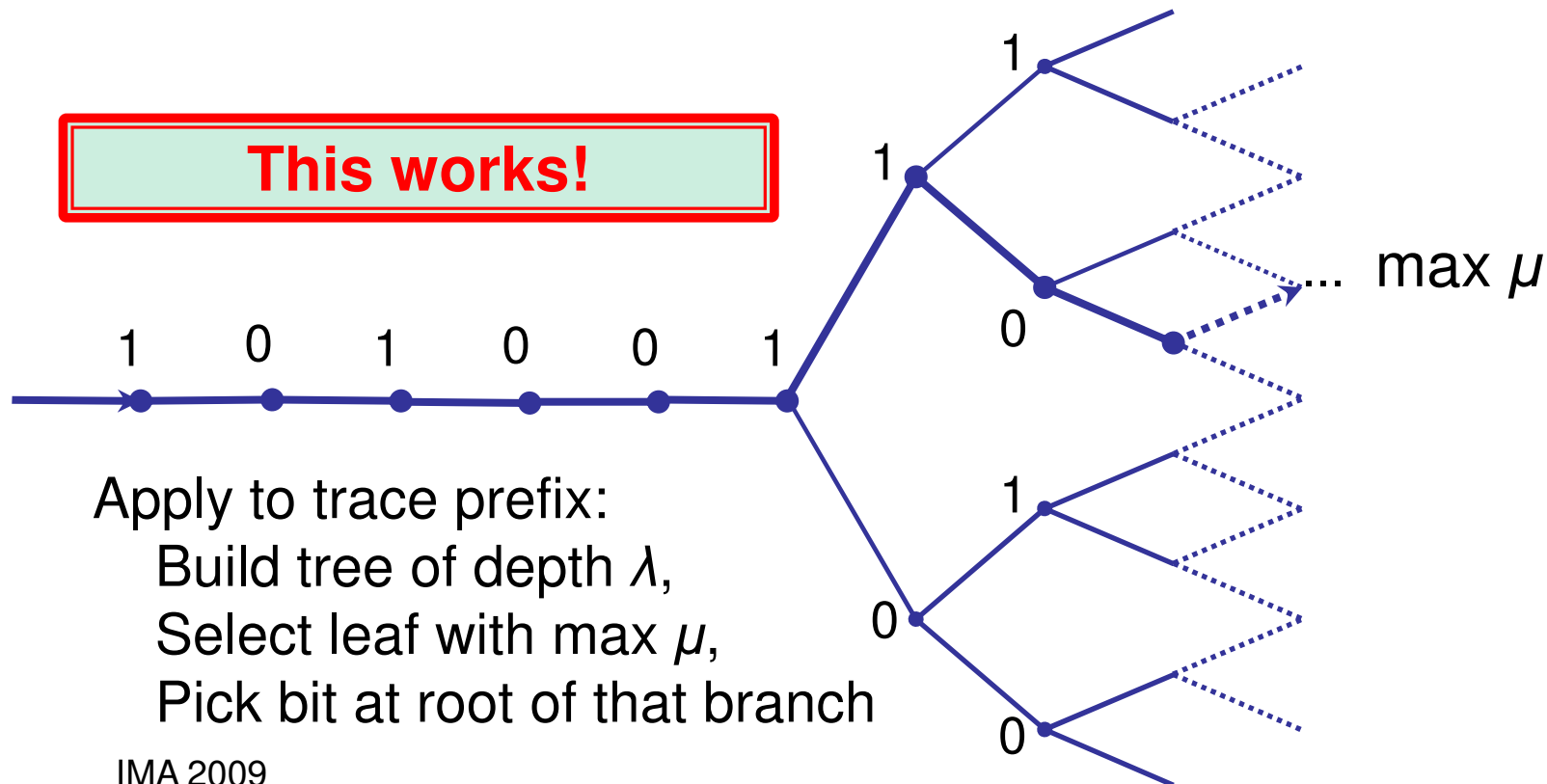
$$\mu(t, K) = \max \{ \mu(t, r) \mid r \text{ is a recoding of } K \}$$

to select the best match recoding of  $K$ .



# The Metric (III)

- Define the **credibility** of key choice  $K$  for a trace  $t$  by
$$\mu(t, K) = \max \{ \mu(t, r) \mid r \text{ is a recoding of } K \}$$
This selects the best match recoding of  $K$ .





# The Metric (IV)

1. Define the *credibility* of a recoding  $r$  for trace  $t$  by

$$\mu(t, r) = \prod_i p_i$$

where  $p_i$  is probability that the  $i$ th operation in  $t$  is the same as the  $i$ th operation for  $r$ .

This should be large for correct interpretation of the trace.

2. Define the *credibility* of a key choice  $K$  for trace  $t$  by

$$\mu(t, K) = \max \{ \mu(t, r) \mid r \text{ is a recoding of } K \}$$

to select the best match recoding of  $K$ .

3. Define the *credibility* of a key  $K$  for trace set  $T$  by

$$\mu(T, K) = \sum_{t \in T} \log(\mu(t, K)) \text{ or } \sum_{t \in T} \mu(t, K)$$

The best fit key maximises this. (The latter is slightly better.)



# Properties

- Traces become aligned correctly (or almost correctly) with key bits/digits by selecting the best fit recoding.
- Summing the metric values for best recodings of each trace provides the averaging that reduces noise and enables the best key bit to be selected.
- Locations for incorrect bits can be determined by looking at the difference in the credibility of the 0- and 1- branches of a node in the tree. A *small* difference means lack of certainty about the decision.
- Key bit positions can be ordered according to this probability of correctness.



# Some Figures

- Take the Ha-Moon 1 recoding.
- Assume a 70% chance of deciding correctly between a square or multiplication from the side channel trace, but *unable* to distinguish the multiplications for  $-1$  and  $+1$ .
- Take typical 192-bit ECC key & *only 5 traces*.
- On average there are only 20.7 bit errors
- In 1.3% of cases there are *no* errors in the 168 bits we are most certain of, leaving just 24 *known* bits to check.
- *It is computationally feasible to correct all errors in these.*



# Conclusion

- Traces from randomised exponentiation algorithms can be aligned effectively to pool weak side channel leakage associated with individual key bits.
- Locations of possible bit errors are identified with ease, making it computationally feasible to correct them.
- Theoretical results on the optimal decision strategy were applied to redesign a previous algorithm for this.
- A more successful algorithm resulted, with sounder basis and better understanding of good parameters to choose.