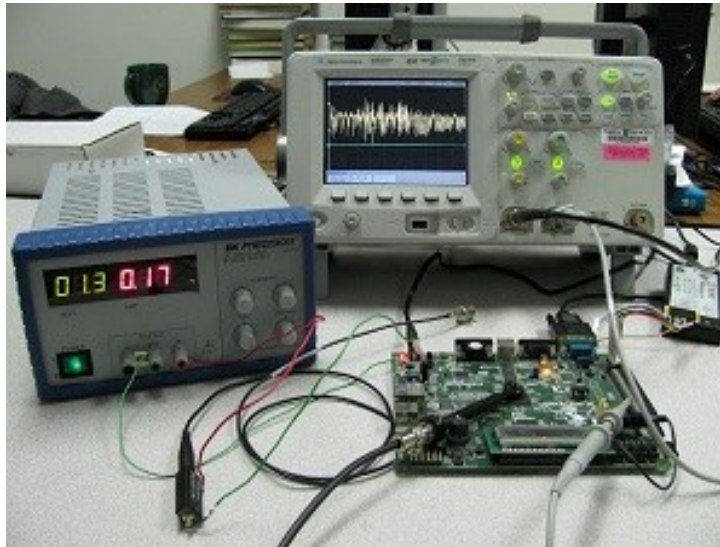


Towards Easy Leakage Certification



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CHES 2016, Santa Barbara, USA

Outline

1. Introduction & motivation
2. (Easy) Leakage certification
3. Experiments
 - Simulations
 - Unprotected software
 - Masked hardware
4. Conclusions

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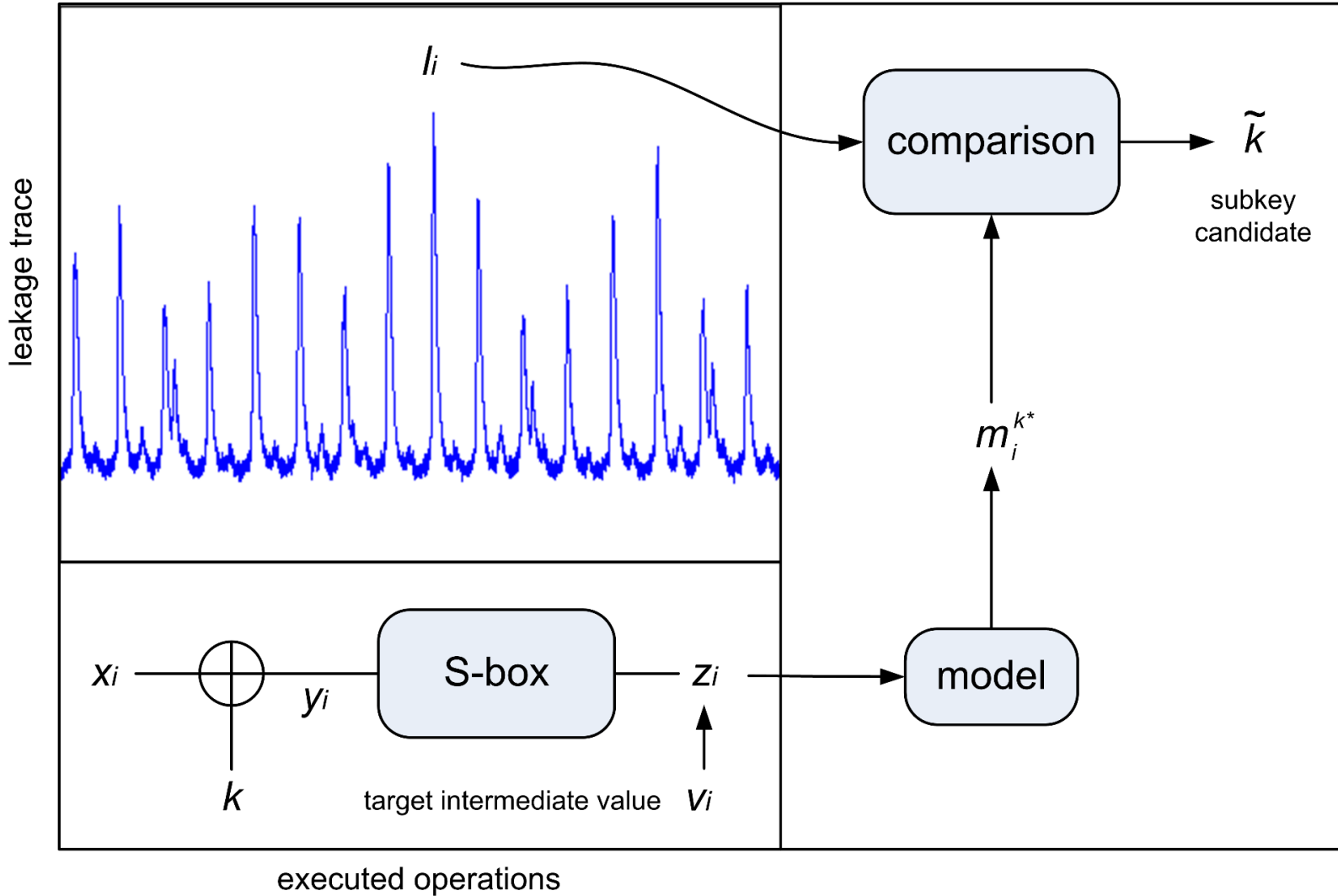
1. Introduction & motivation

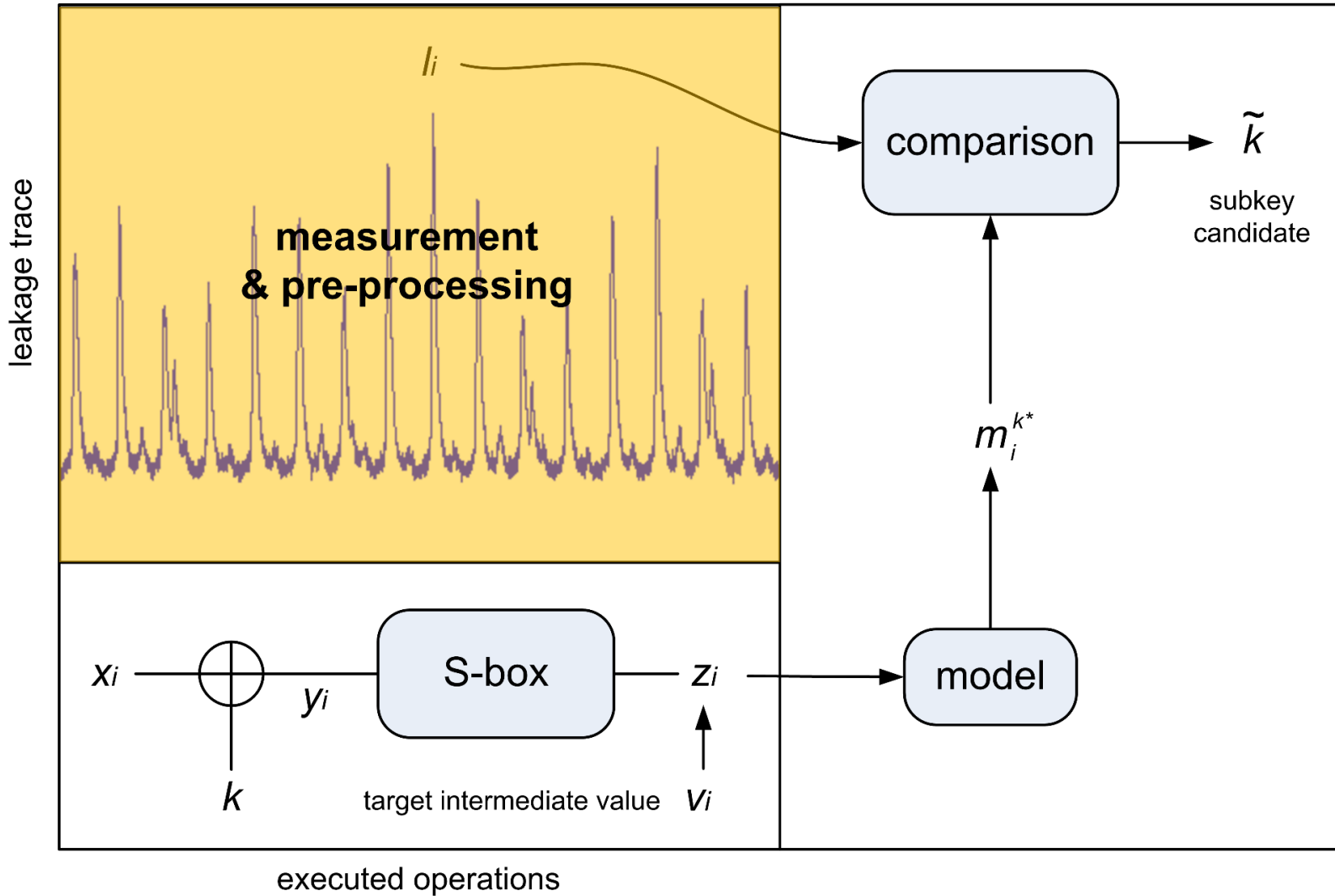
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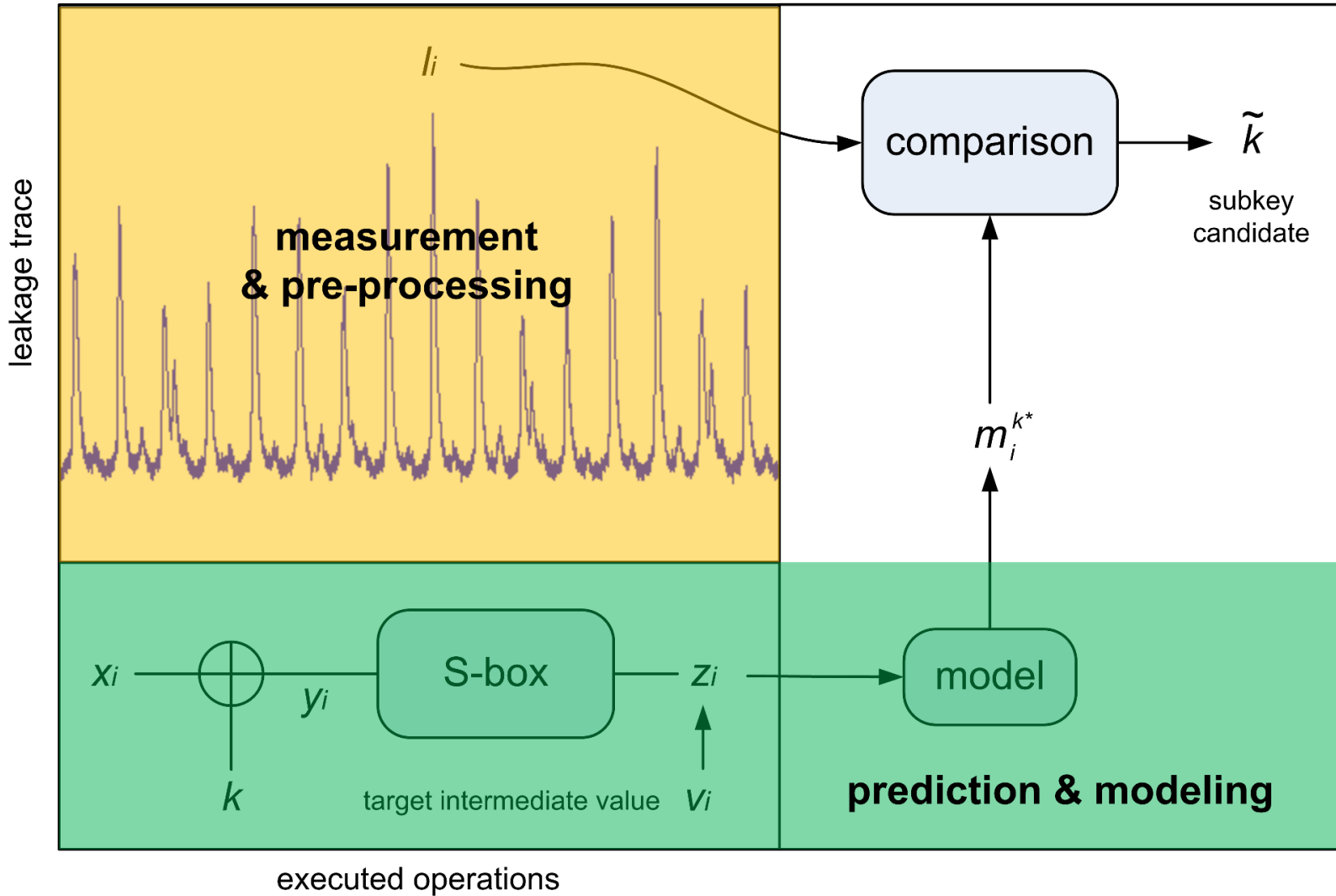
3. Experiments

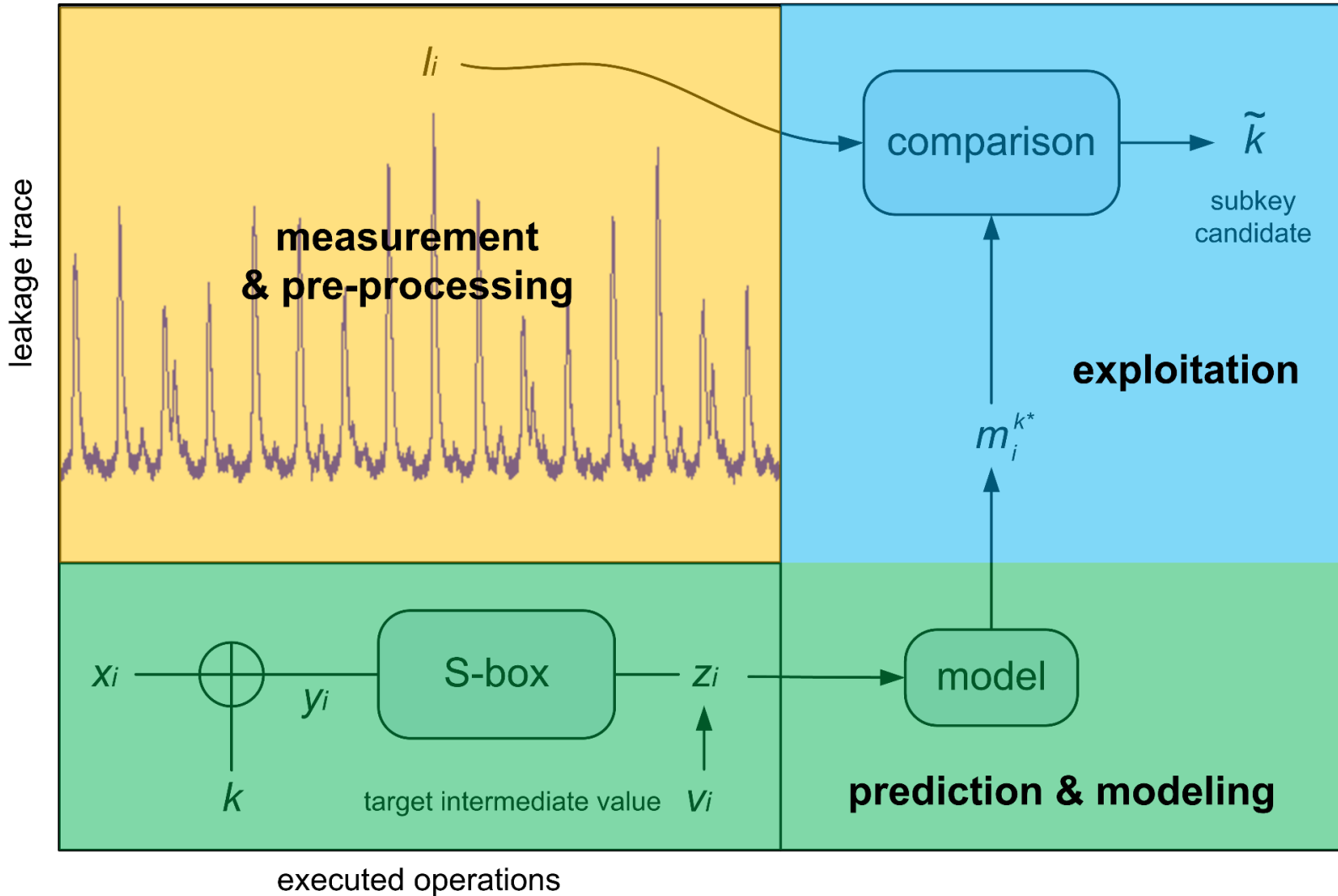
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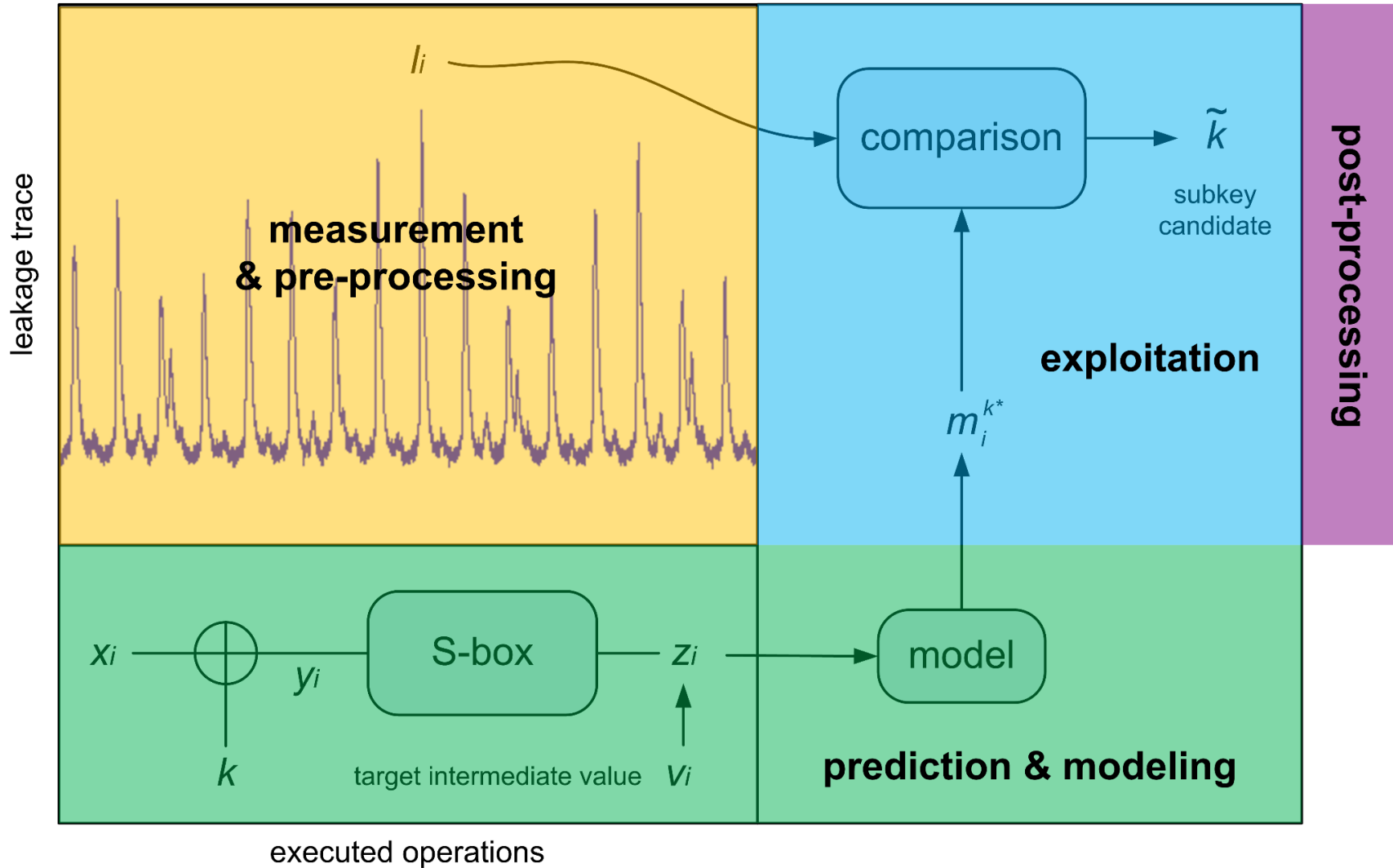
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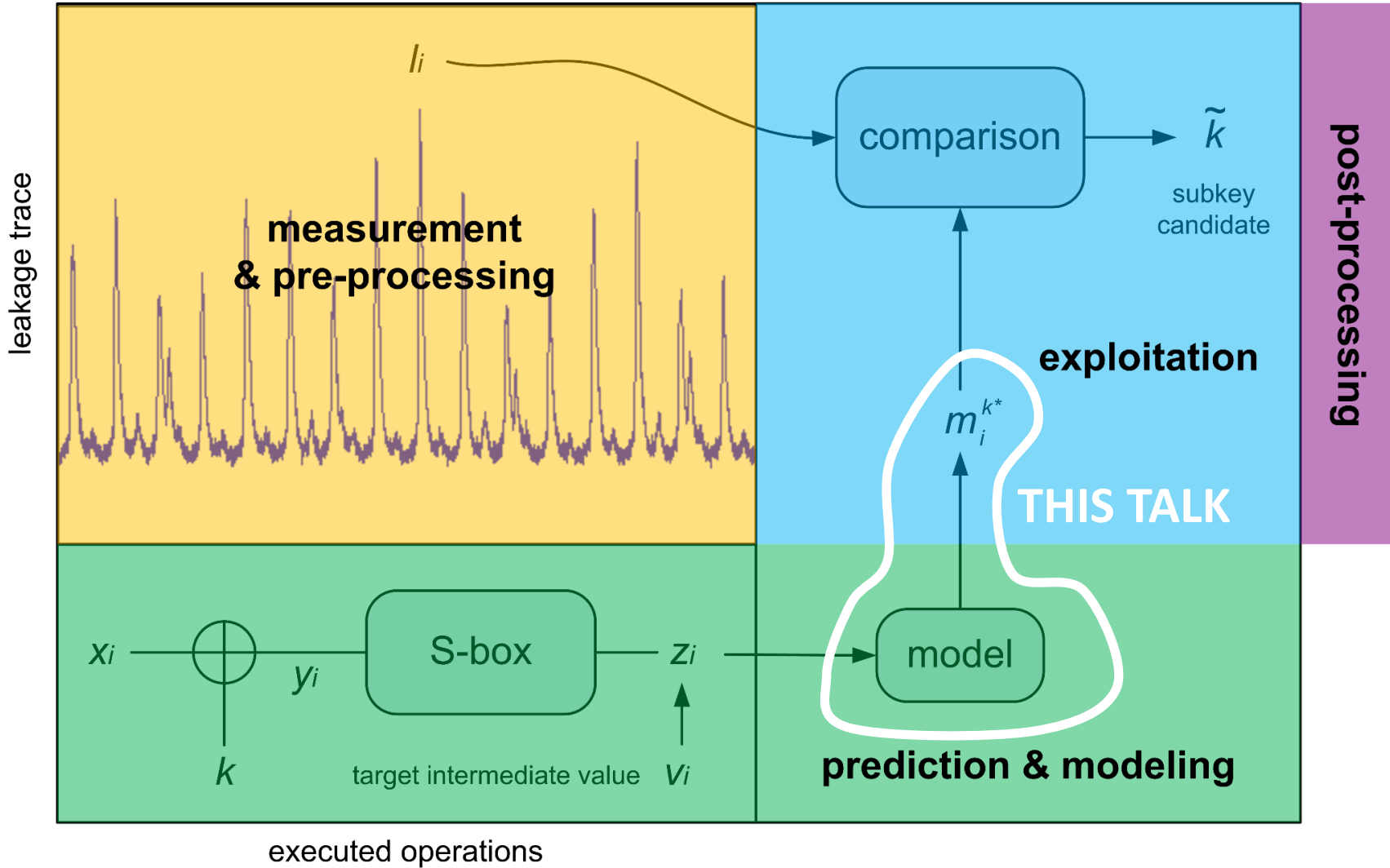






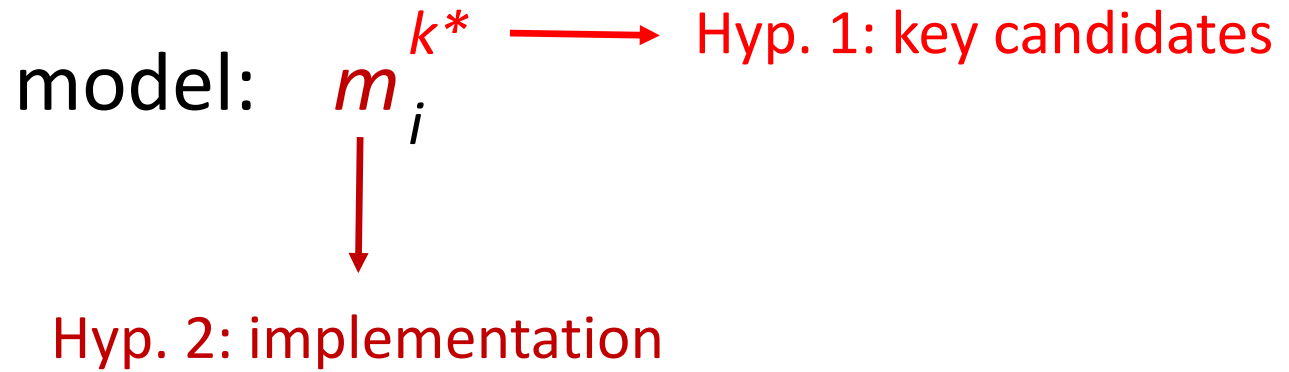






model: $m_i^{k^*}$

model: $m_i^{k^*}$ \longrightarrow Hyp. 1: key candidates



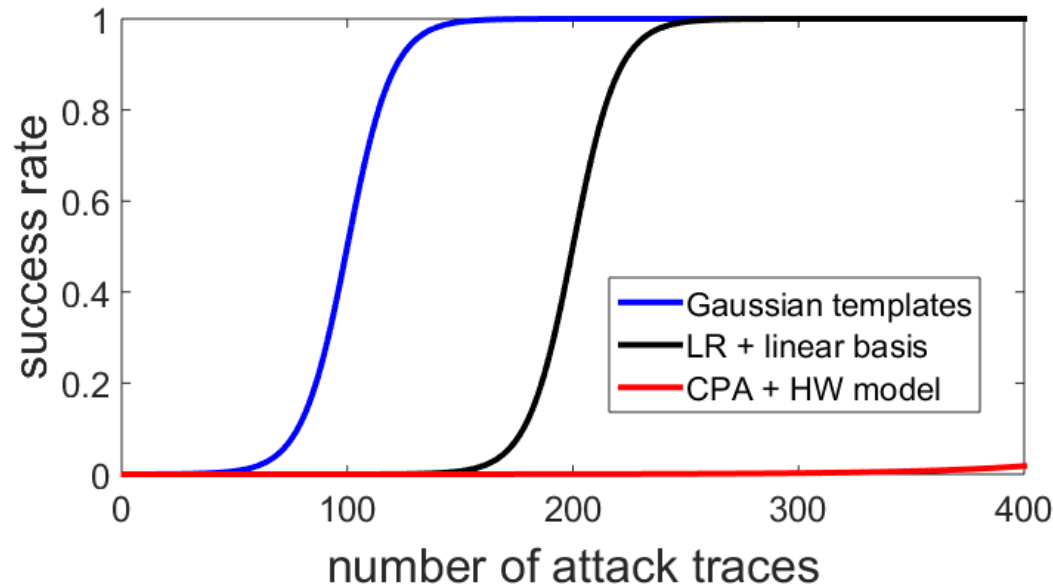
- For the key candidates, we try them all

- But it is impossible to try all models! [W12]

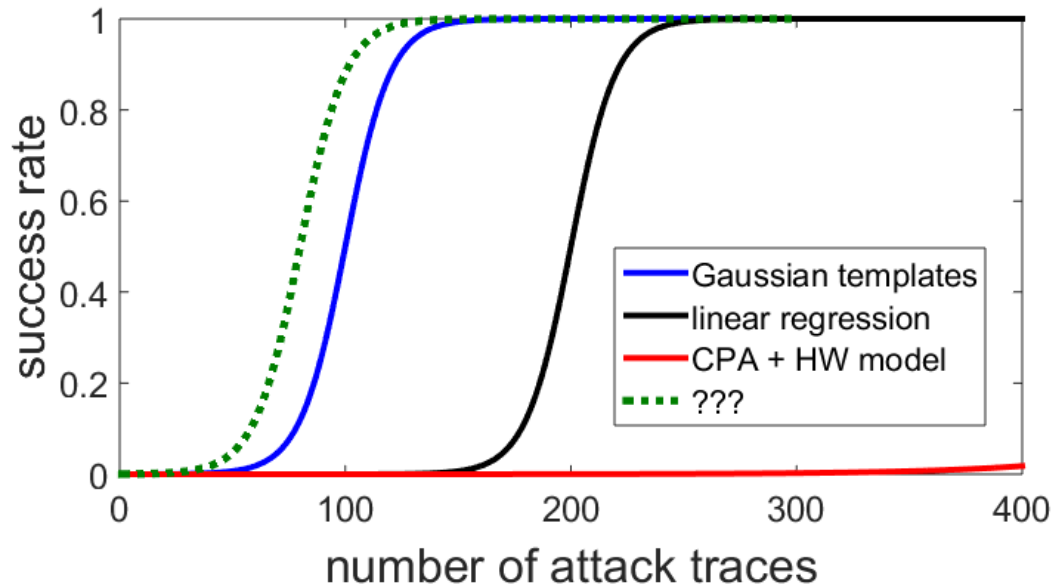
⇒ How to be sure the model is “good enough”?

- Does it really happen in practice?

- Does it really happen in practice?



- Each time a model performs better than another



⇒ How to be sure the model is “good enough”?

- A model is optimal if $\hat{\text{Pr}}_{model} [l|k] = \text{Pr}_{chip} [l|k]$

⇒ Theory would say it is ε -close to optimal if

$$\text{SD}(\hat{\text{Pr}}_{\text{model}} [l|k], \text{Pr}_{\text{chip}} [l|k]) < \varepsilon$$

- (with SD a statistical distance)

- Convenient since ε would quantify the loss
 - That could be reported in SR bounds [DFS15]

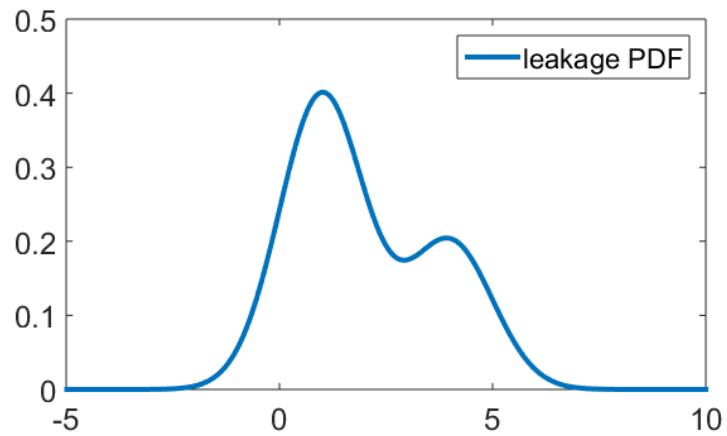
- Problem: $\Pr_{chip} [l|k]$ is unknown

Outline

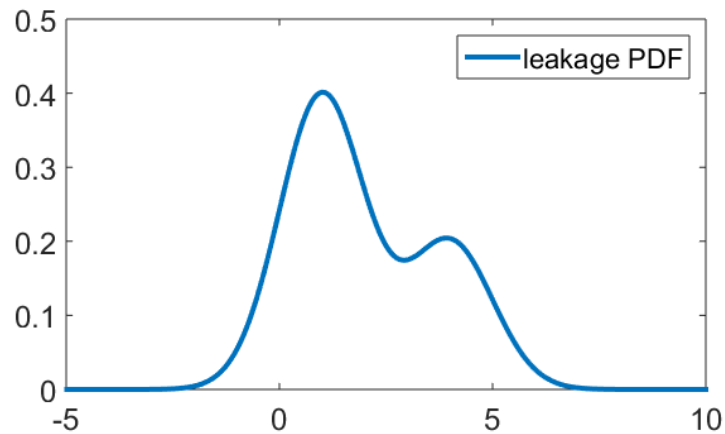
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- Distinguish estimation & assumption errors
 - Recall estimation errors decrease with # meas.

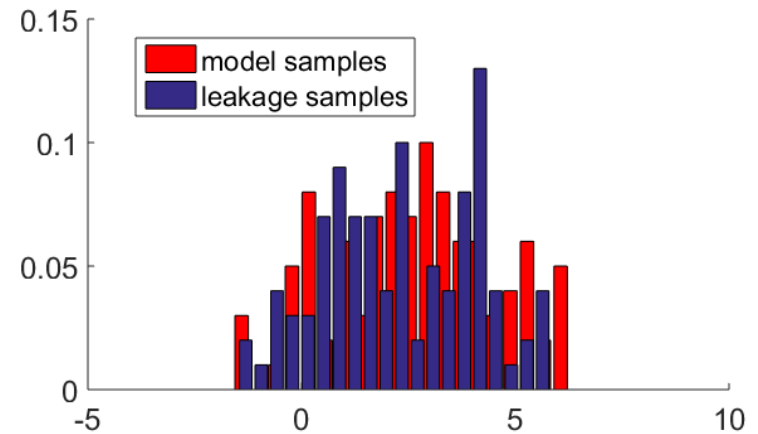
- Example:



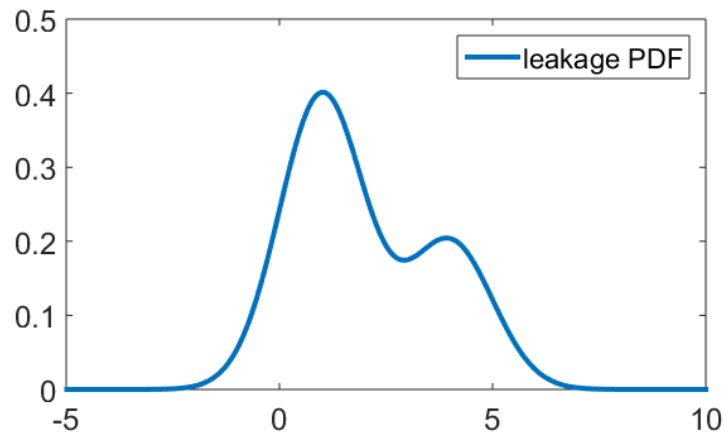
- Example:



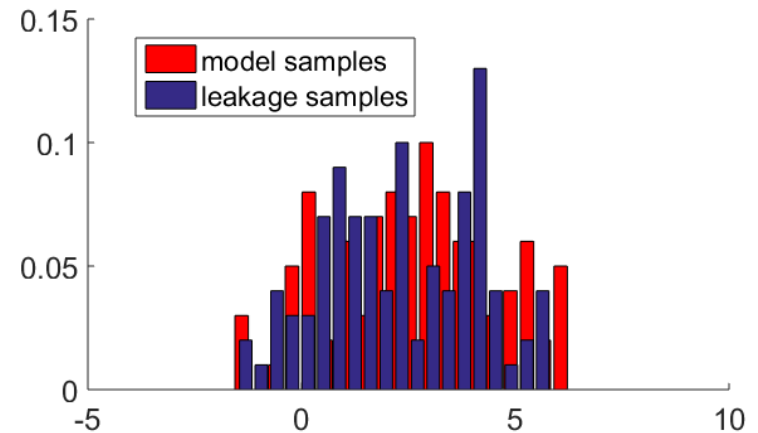
No samples



- Example:



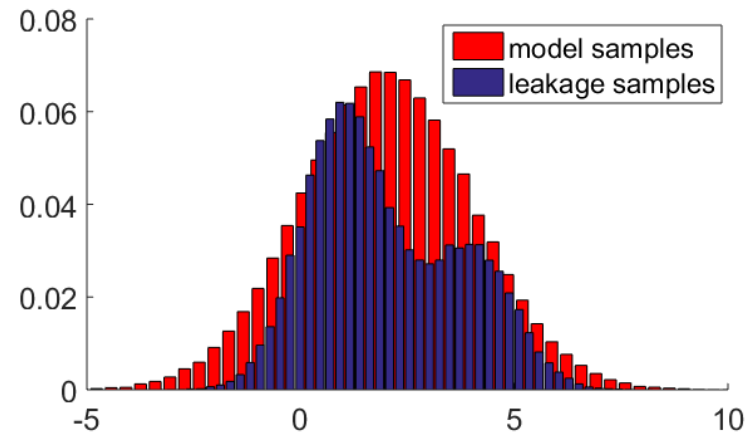
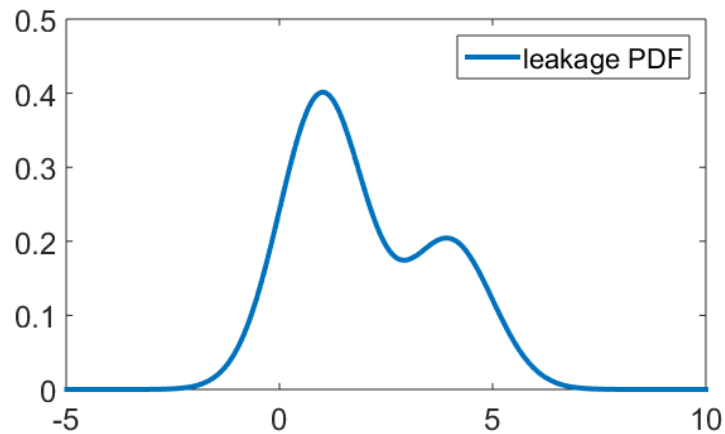
estimation errors dominate



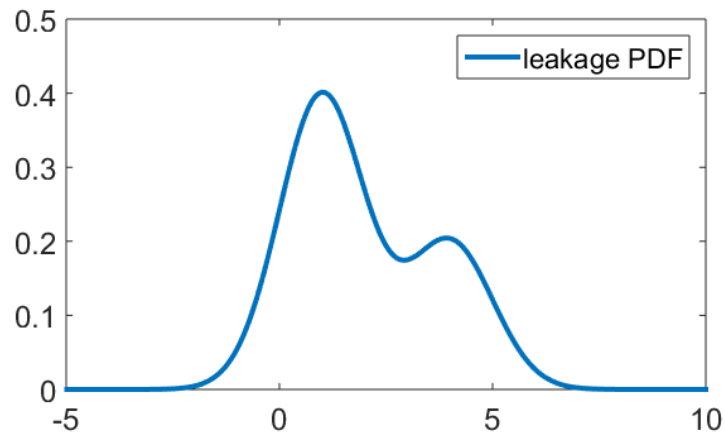
⇒ need to measure more

- Example:

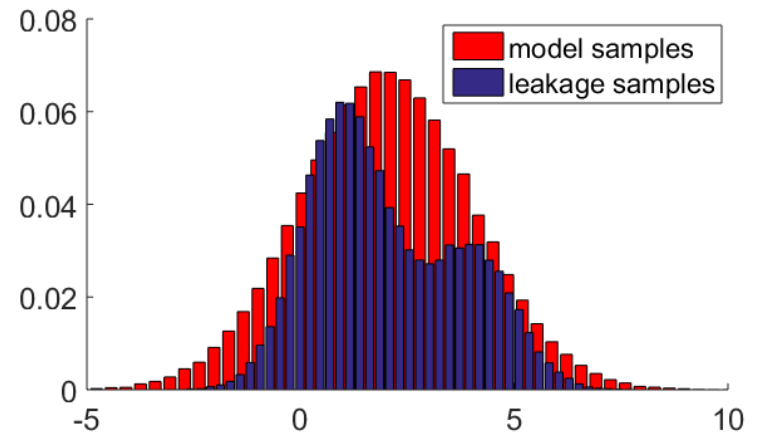
$N_1 > N_0$ samples



- Example:



assumption errors dominate



⇒ need another model

\Rightarrow good enough model: *ass. err* \ll *est. err.* given N

- Test the hypothesis that

$$\hat{\Pr}_{model} [l|k] \stackrel{N}{=} \Pr_{chip} [l|k]$$

- Taking advantage of cross-validation



modeling samples



test samples



- Taking advantage of cross-validation



modeling samples



test samples



- Taking advantage of cross-validation



modeling samples



test samples



- Taking advantage of cross-validation



modeling samples



test samples



- Output a p-value $p(N)$
 - Small p 's indicate hyp. is likely incorrect

- Output a p-value $p(N)$  Eval. lab. limit

- Main drawback: cost (of sampling distributions)

- Compare moments (rather than distributions)

1. $\hat{M}_d \stackrel{N}{\leftarrow} \hat{\text{Pr}}_{model} [l|k]$

2. $\tilde{M}_d \stackrel{N}{\leftarrow} \text{Pr}_{chip} [l|k]$

3. Test equality

$$\hat{M}_d = \tilde{M}_d$$

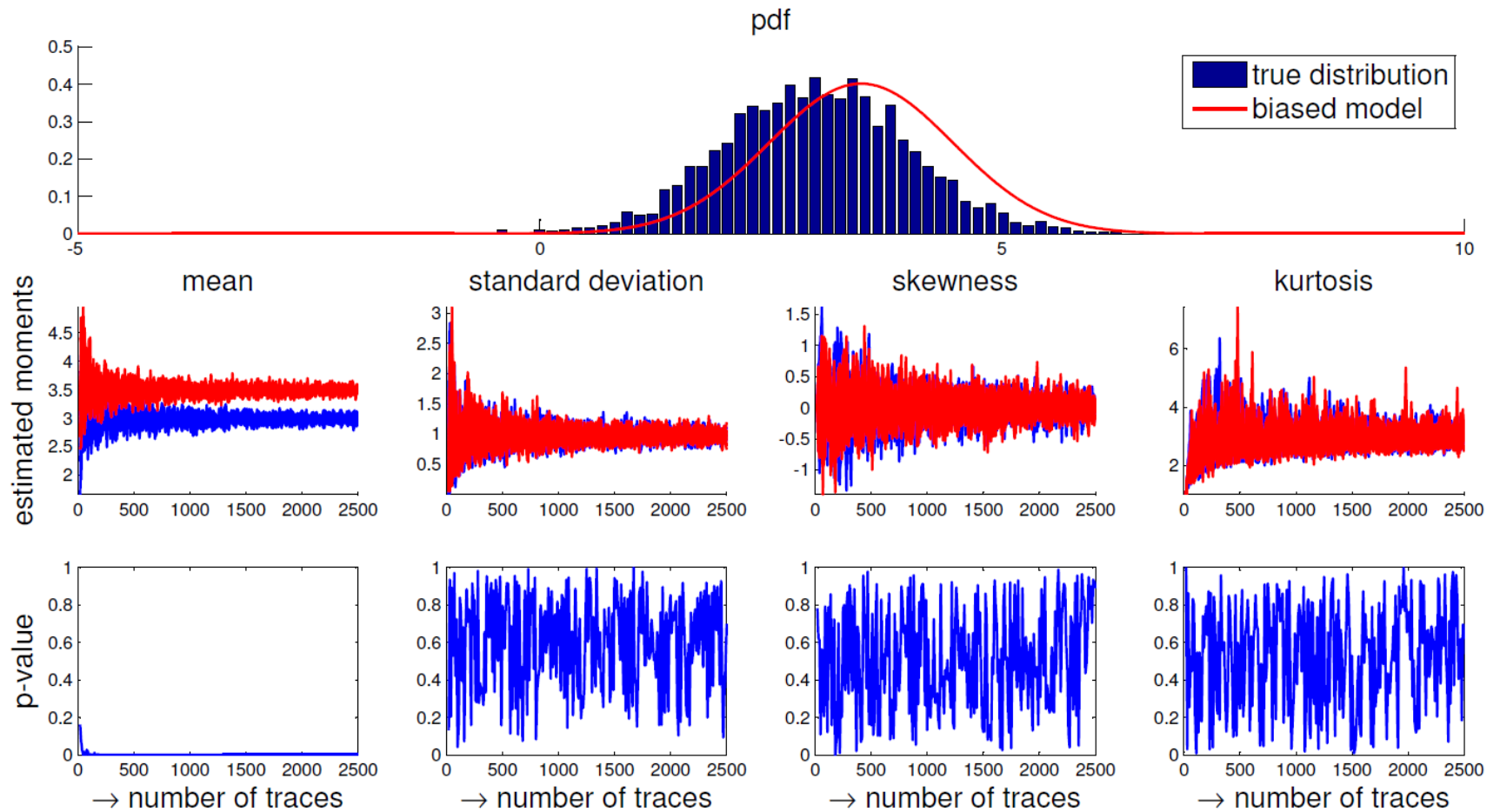
- + Can be done with simple univariate tests
 - e.g., T-test (assuming \hat{M}_d, \tilde{M}_d are Gaussian)

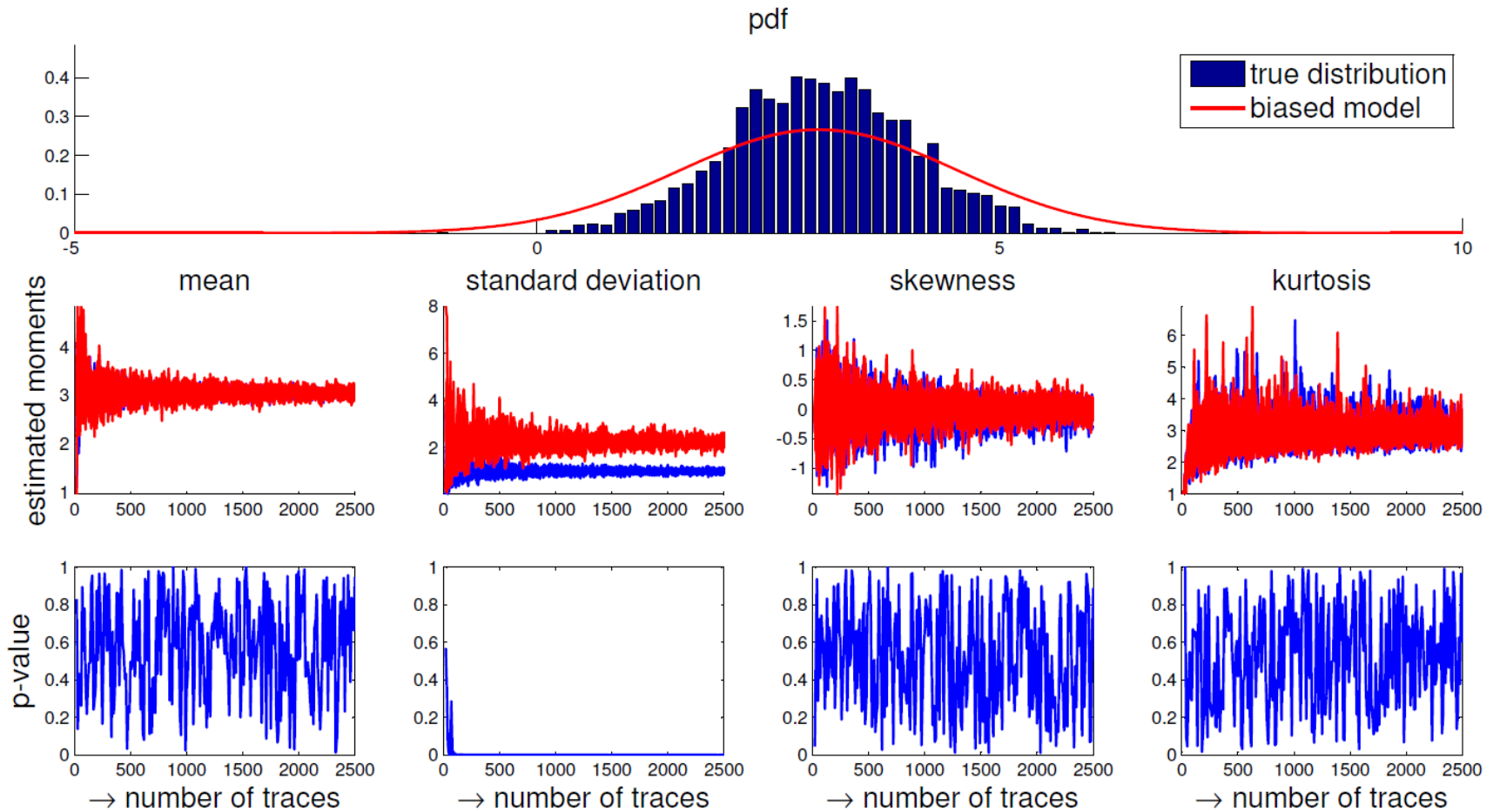
- Is it theoretically sound? No!

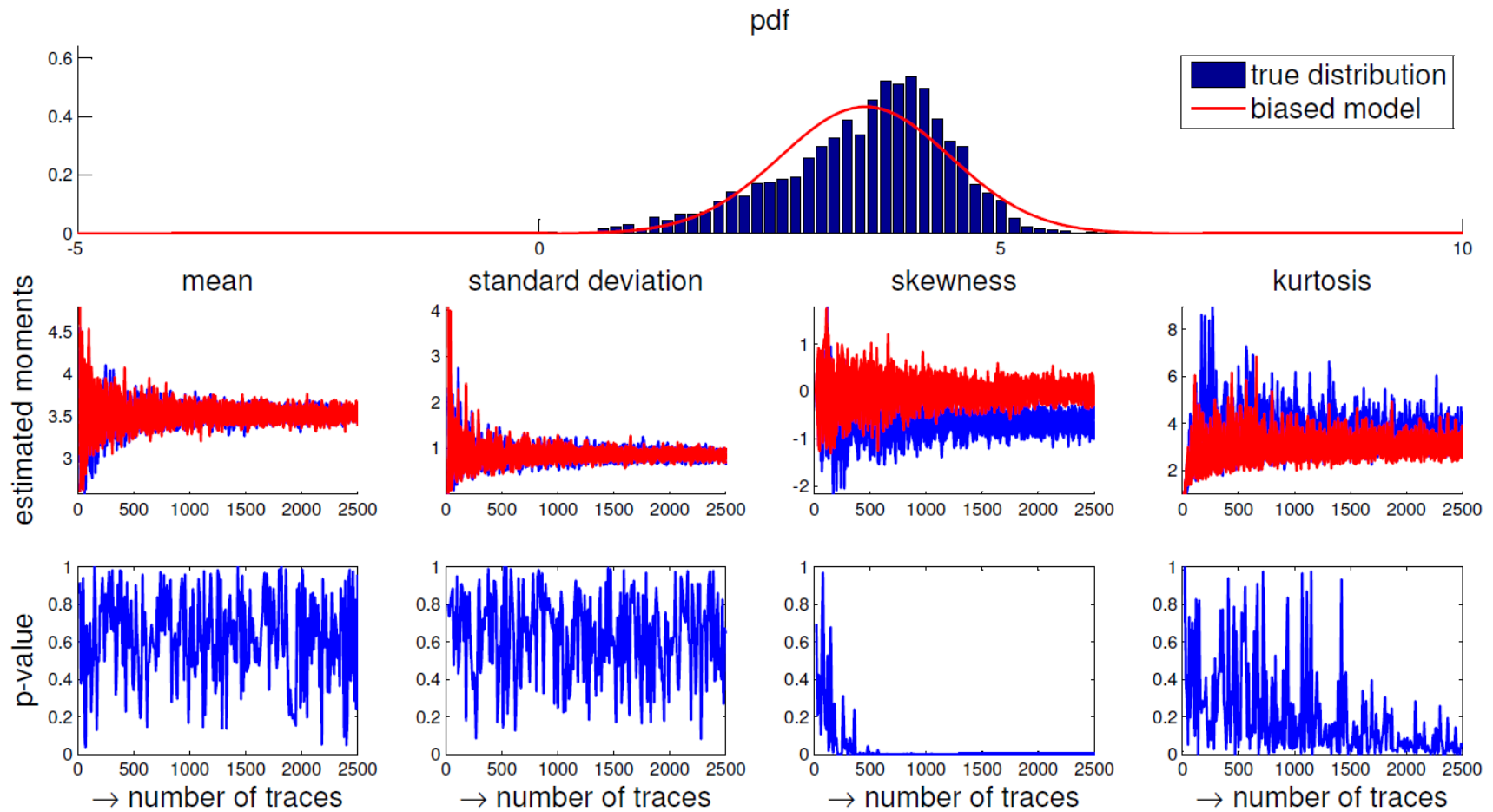
- Is it theoretically sound? No!
 - But counterexamples are involved
 - & SCA literature frequently does it
 - Leakage detection, HO attacks, ... [SM15]

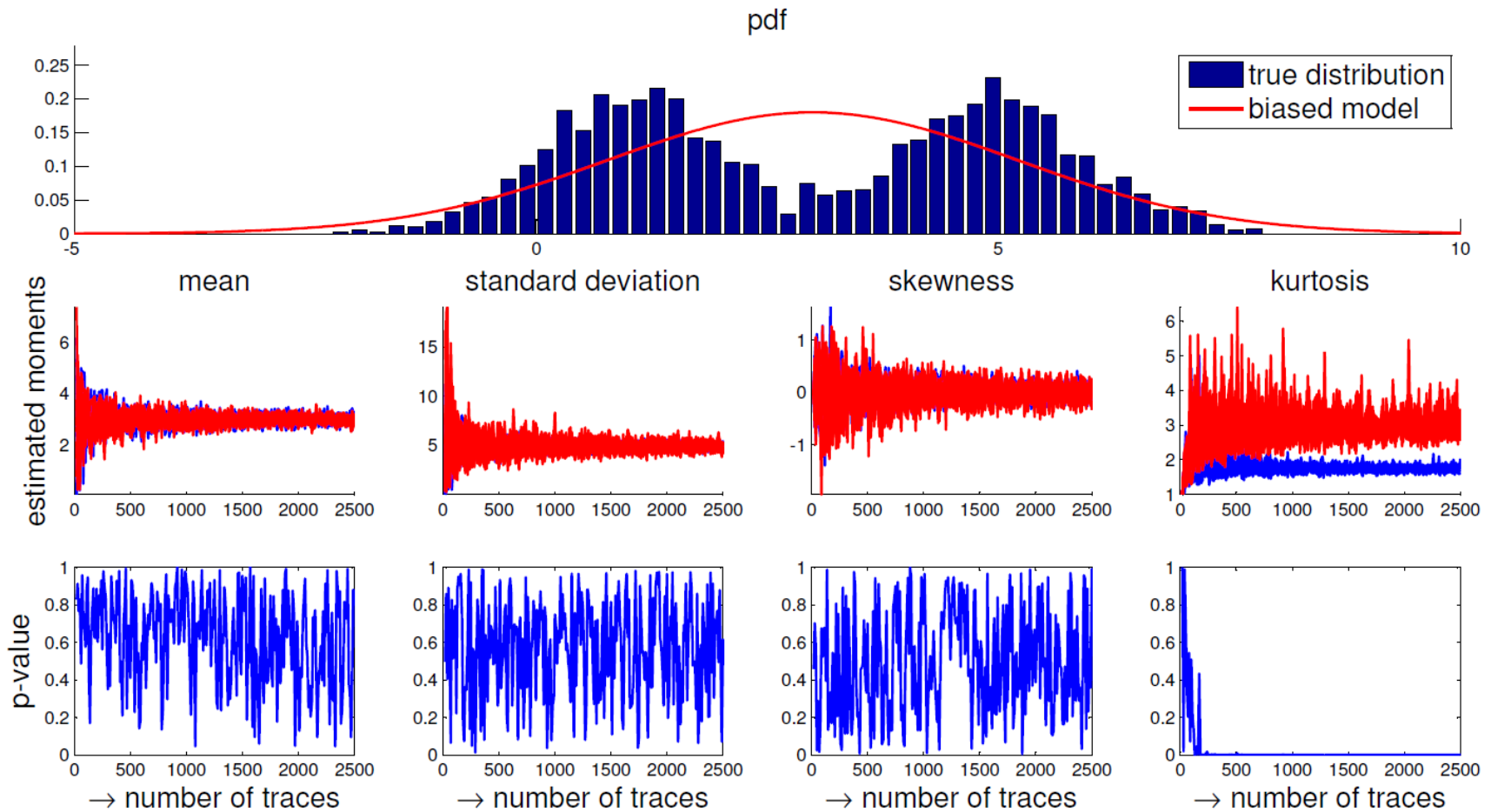
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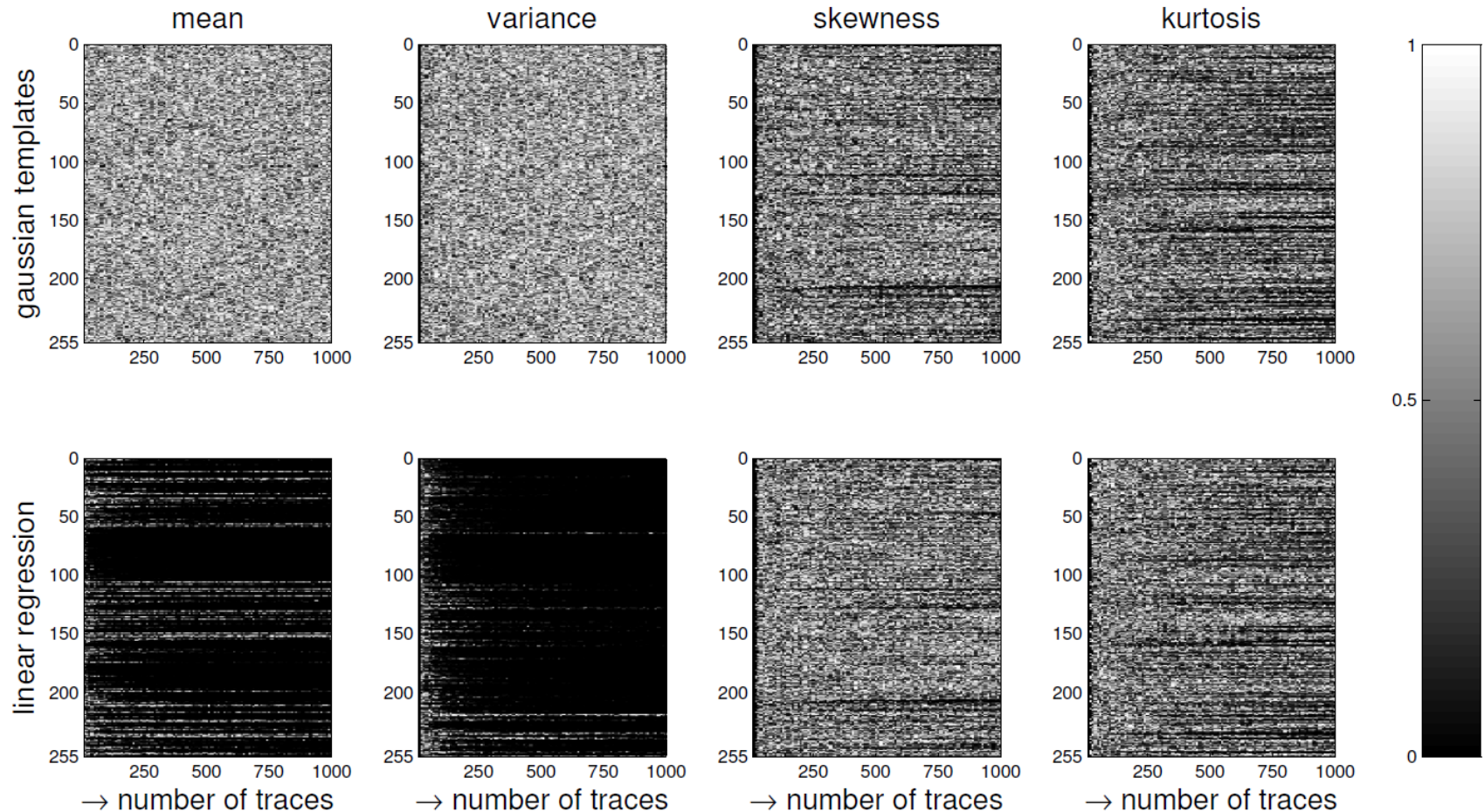
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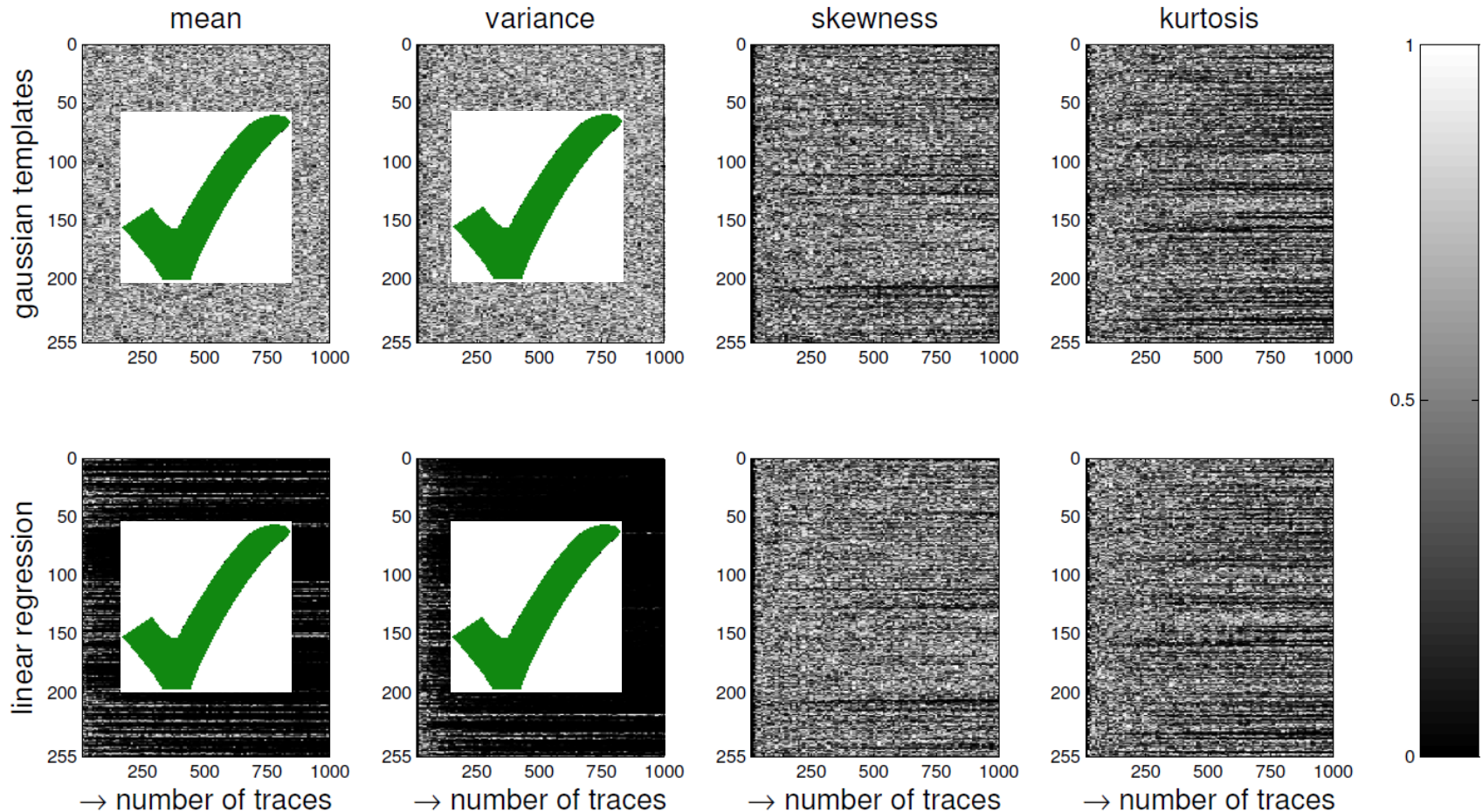
- Repeating the Eurocrypt 2014 case study

- Unprotected AES implementation, Atmel AVR

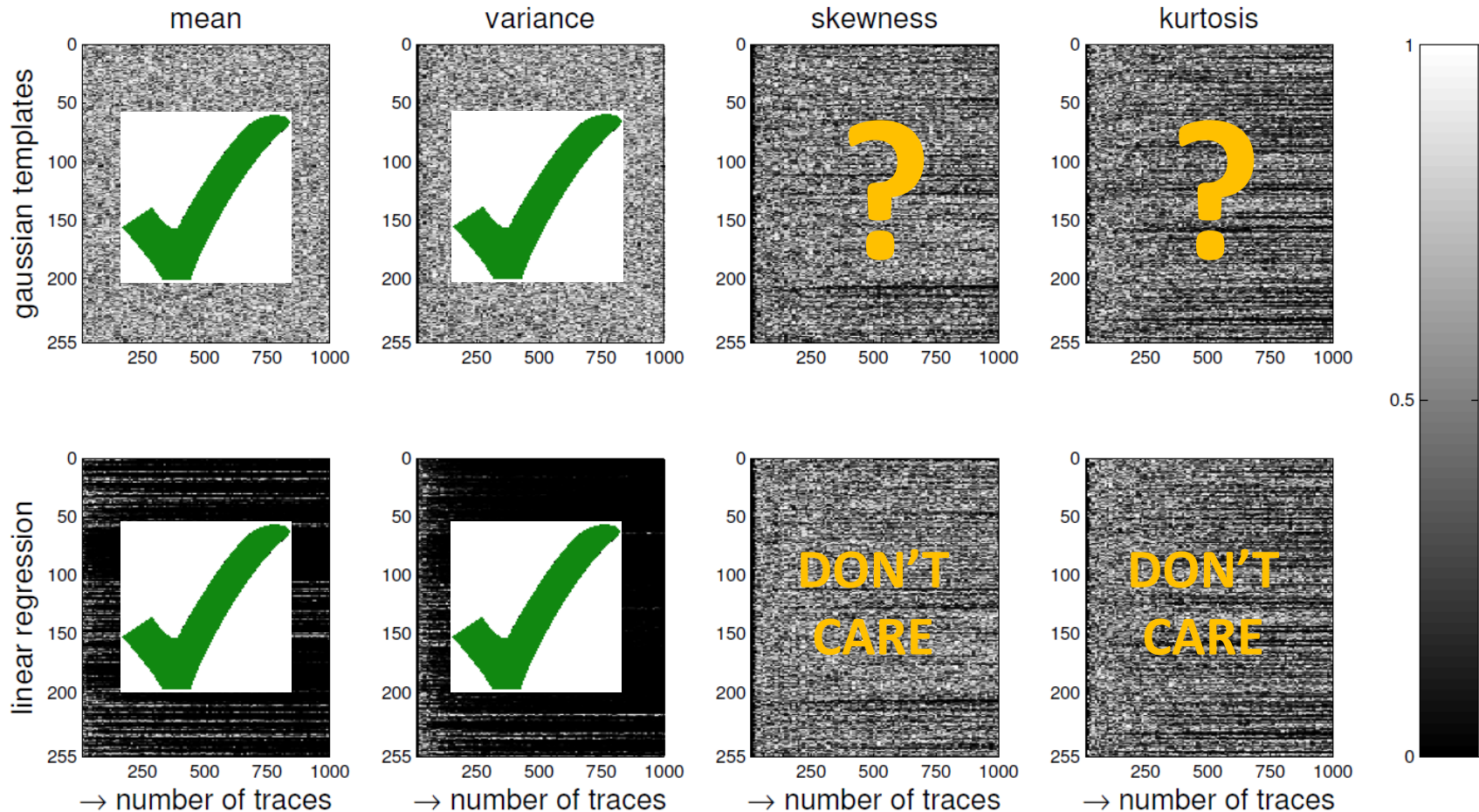
- Unprotected AES implementation, Atmel AVR



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- Unprotected AES implementation, Atmel AVR



- Eurocrypt 2014: no errors detected with up to 256x1000 measurements & Gaussian template
- CHES 2016: small errors in \tilde{M}_3 and \tilde{M}_4

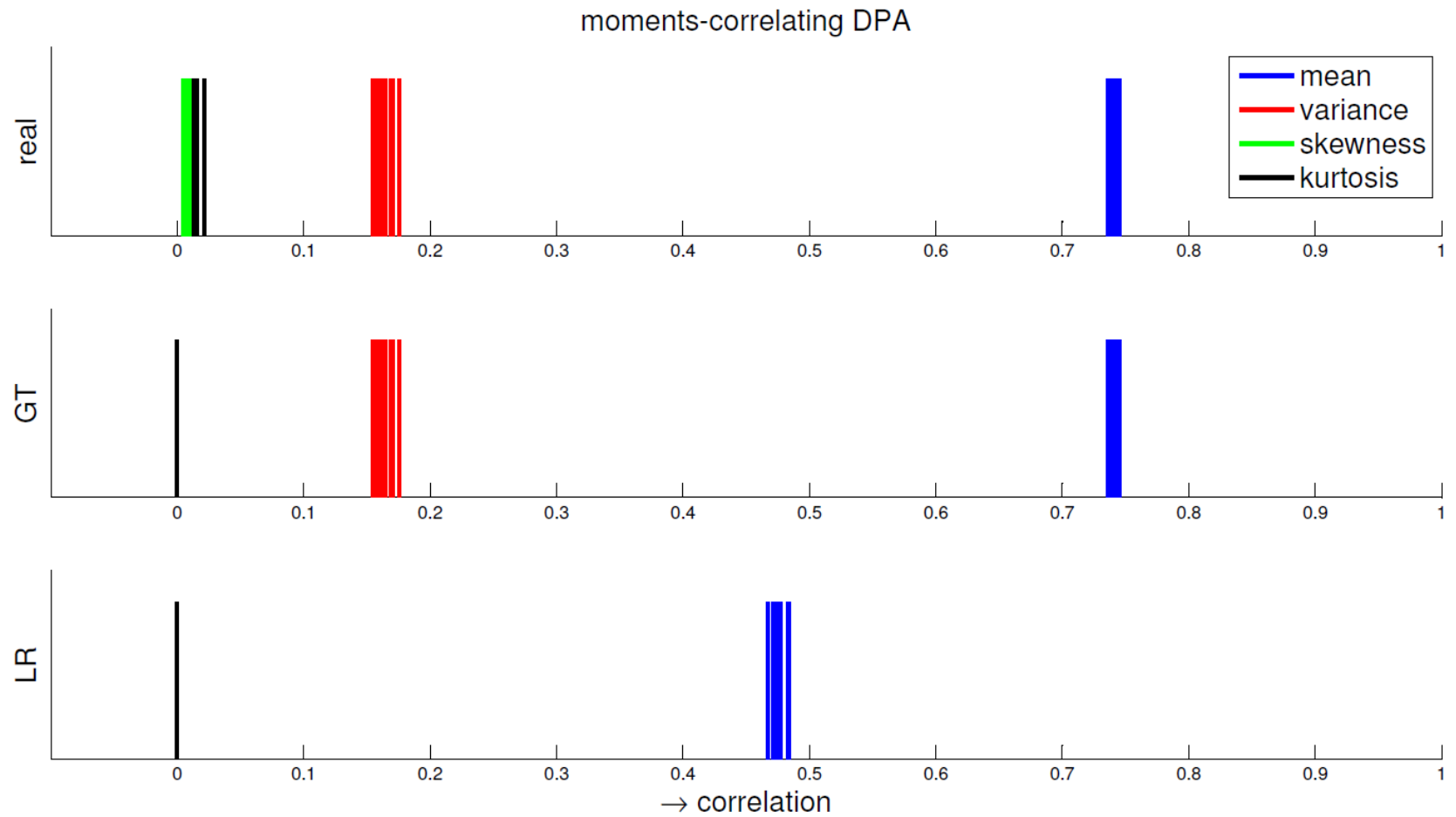
⇒ *Is there an inconsistency in our results?*

⇒ Do these errors lead to significant information loss

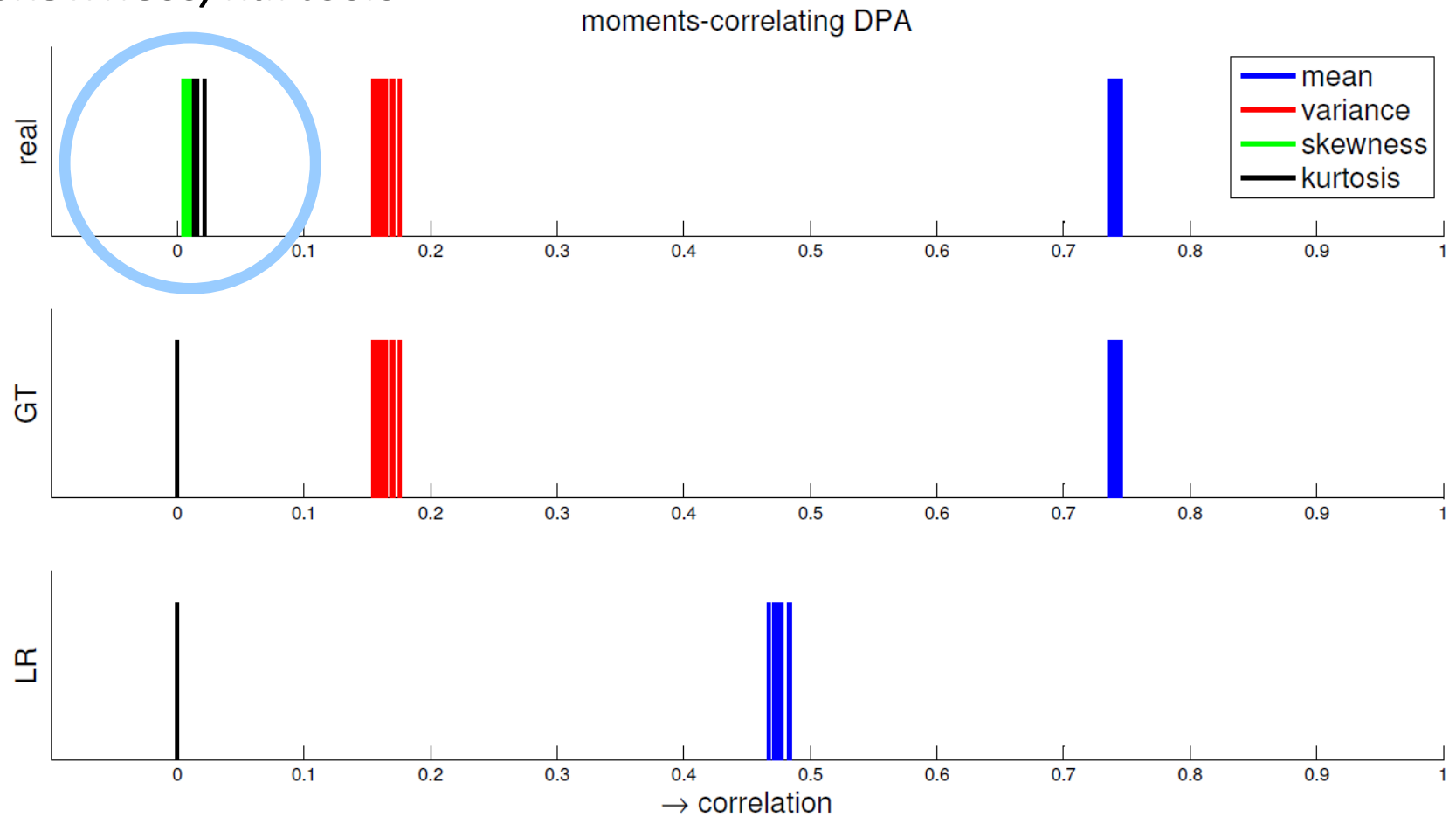
- Additional test: Moments-Correlating DPA [MS14]

$$\text{MPC-DPA}(d) = \hat{\rho}(\hat{M}_d, l^d)$$

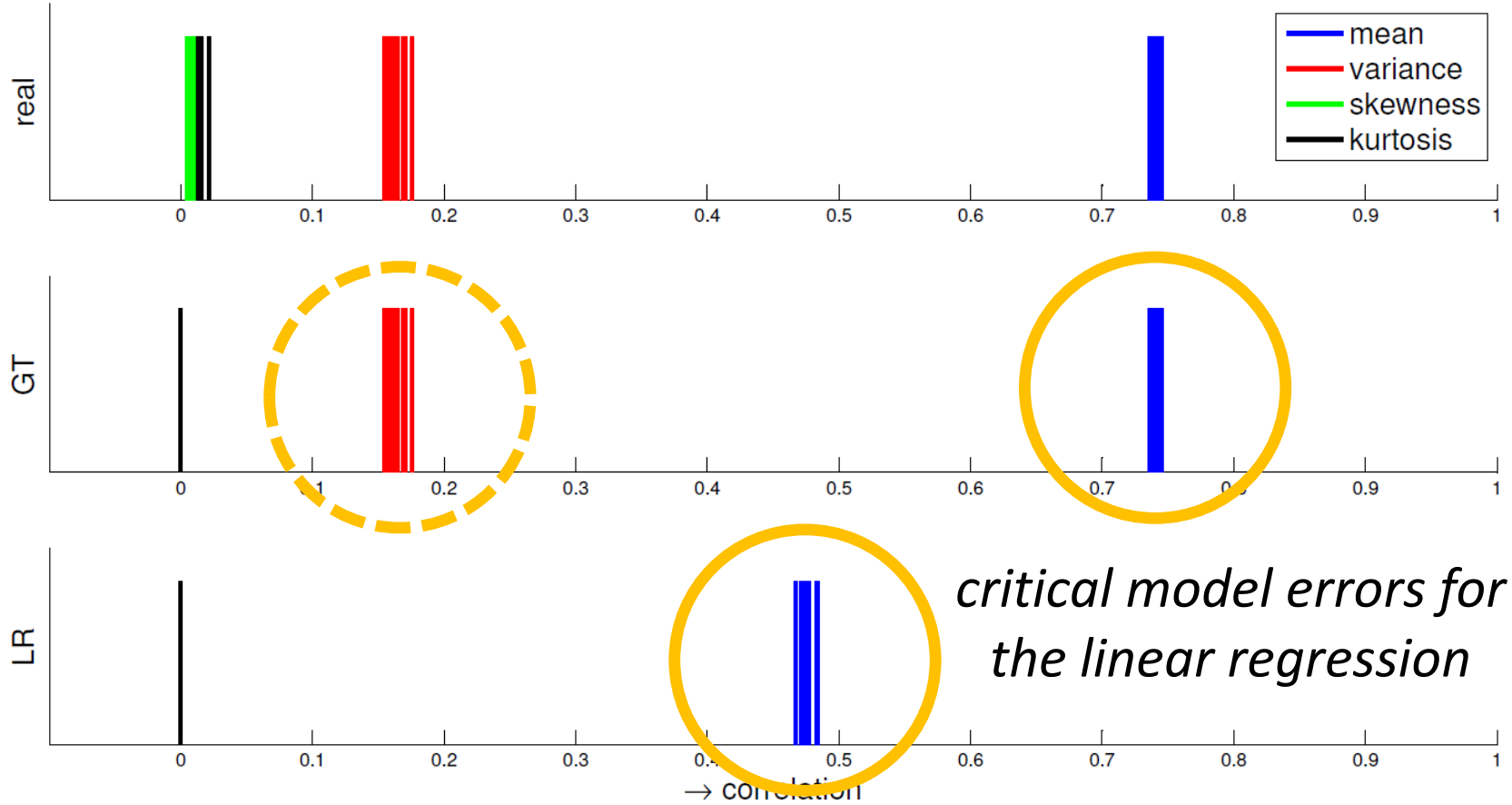
- Metric intuition: $N_s = \frac{c}{\hat{\rho}(\hat{M}_d, l^d)^2}$



*little information in
skewness/kurtosis*



moments-correlating DPA

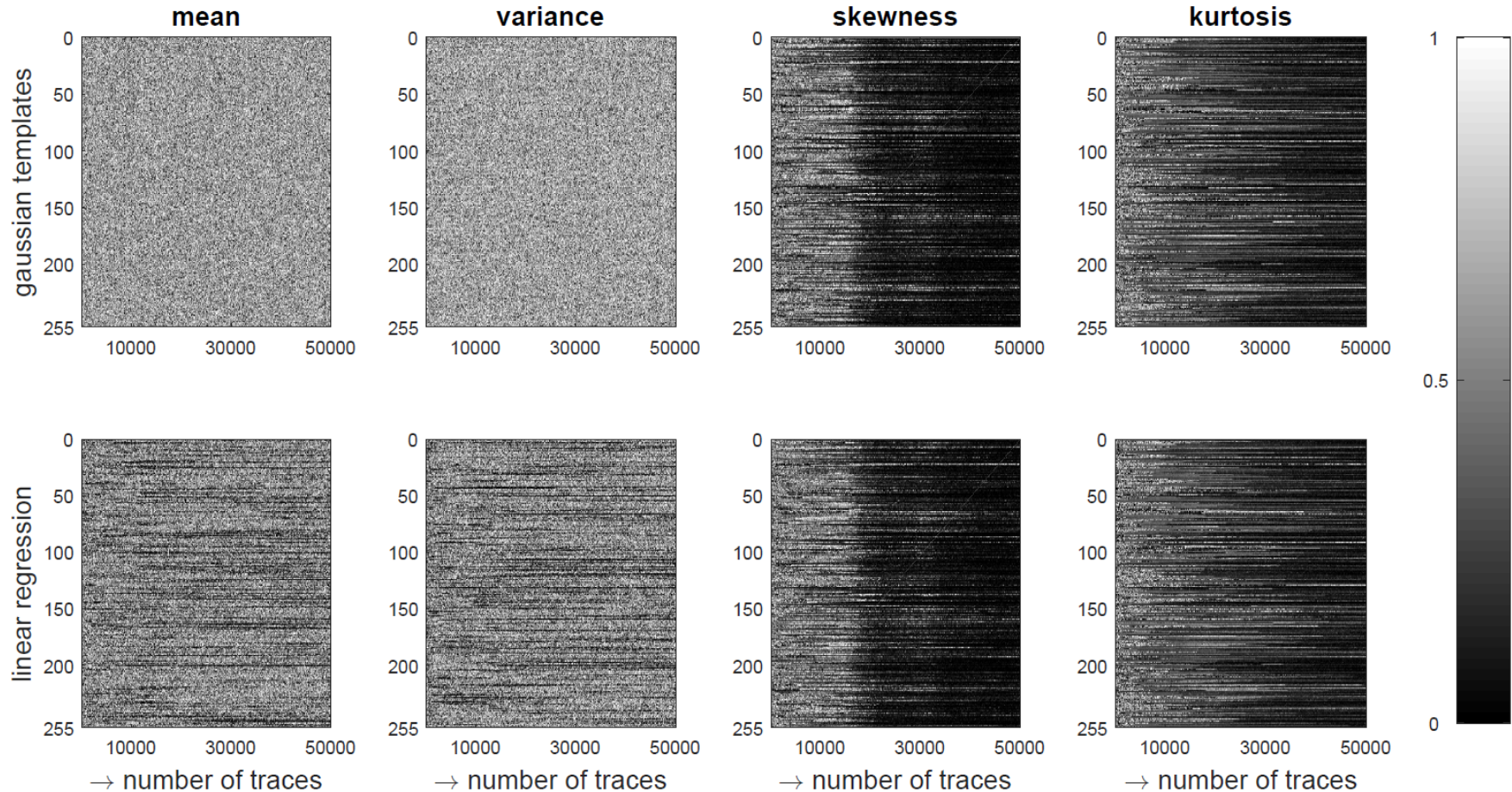


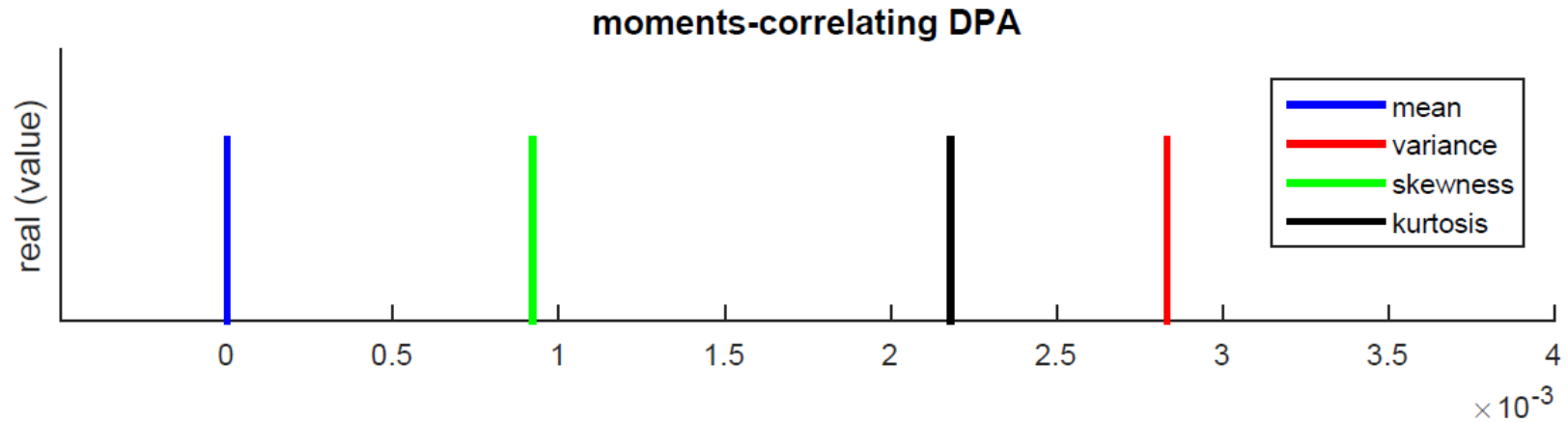
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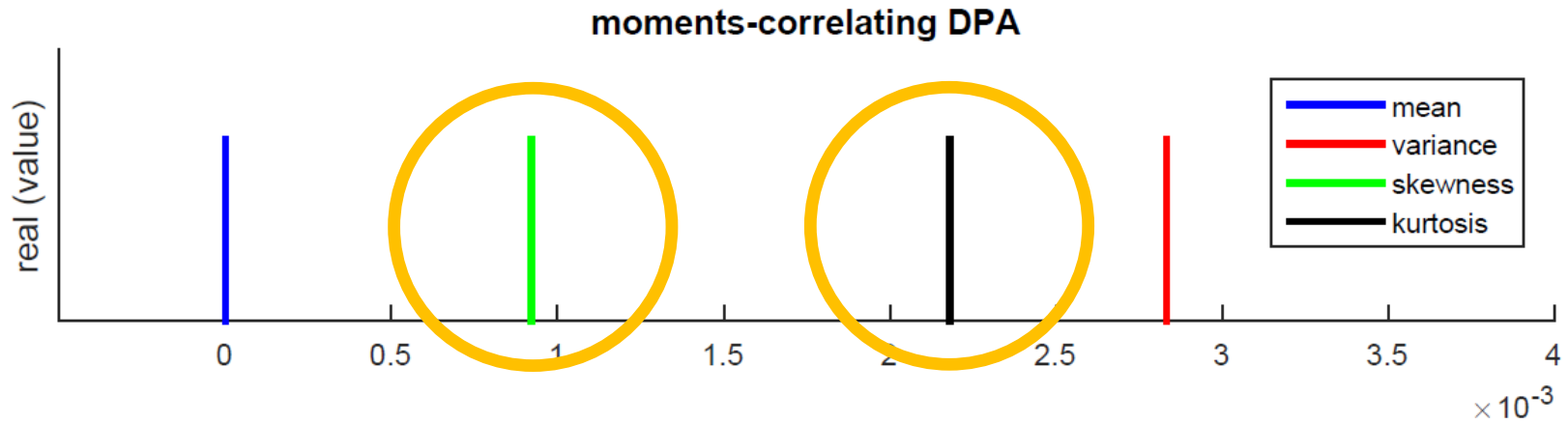
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- 1st-order secure threshold implementation [P+11]

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critical model errors for the Gaussian templates

- As expected since GT capture only 2 moments
⇒ More complex models needed in this case [S+16]

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- Less formal but more efficient/intuitive tool

- Less formal but more efficient/intuitive tool
 - \approx as efficient as profiled CPA
 - (But still benefits from POI detection)
 - Provides hints about the information losses

- Prototype open source code:

<http://perso.uclouvain.be/fstandae/PUBLIS/171.zip>

- Open problems: how to efficiently deal with multivariate & higher-order distributions
- Moment- vs. distribution-based evaluations?

PS. No assumption errors if non-parametric estimations

