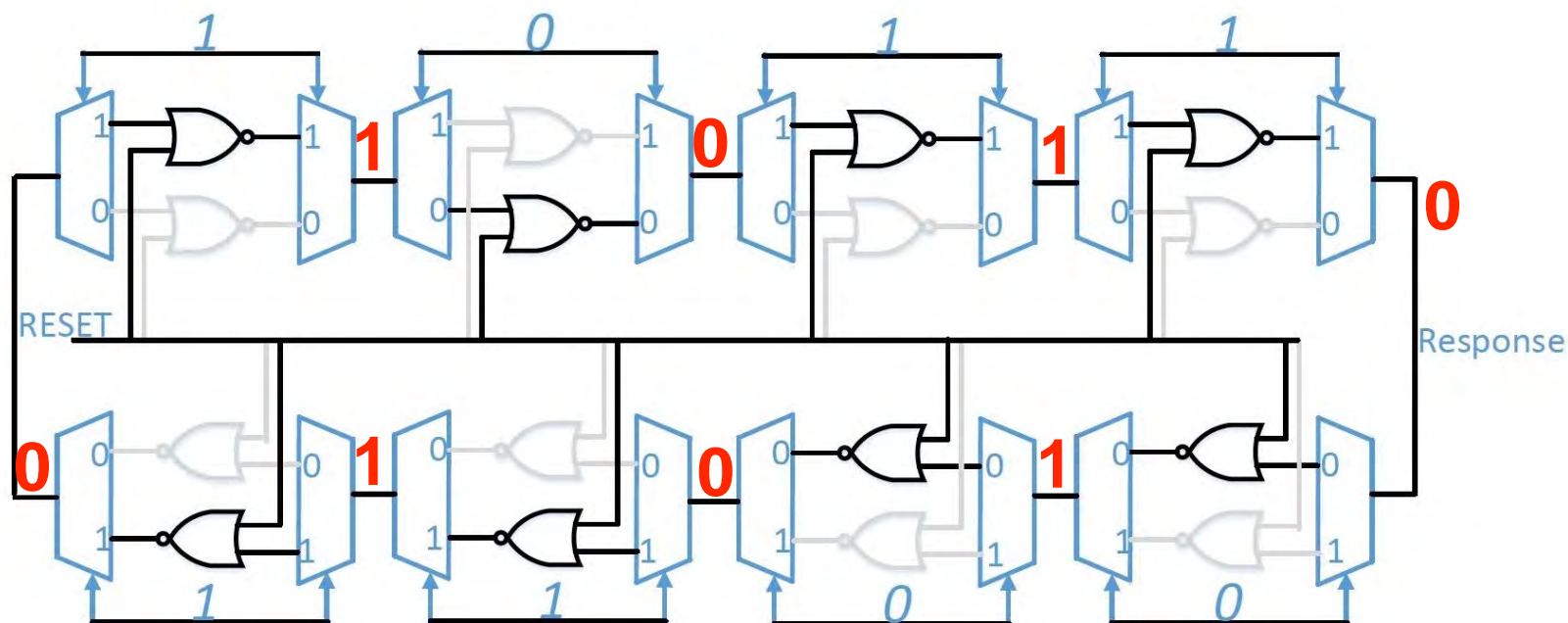


Example

- Challenge bits select weights, stage index determines signs
- Response tells whether sum is negative or positive
- Additive delay model (like Arbiter PUF)

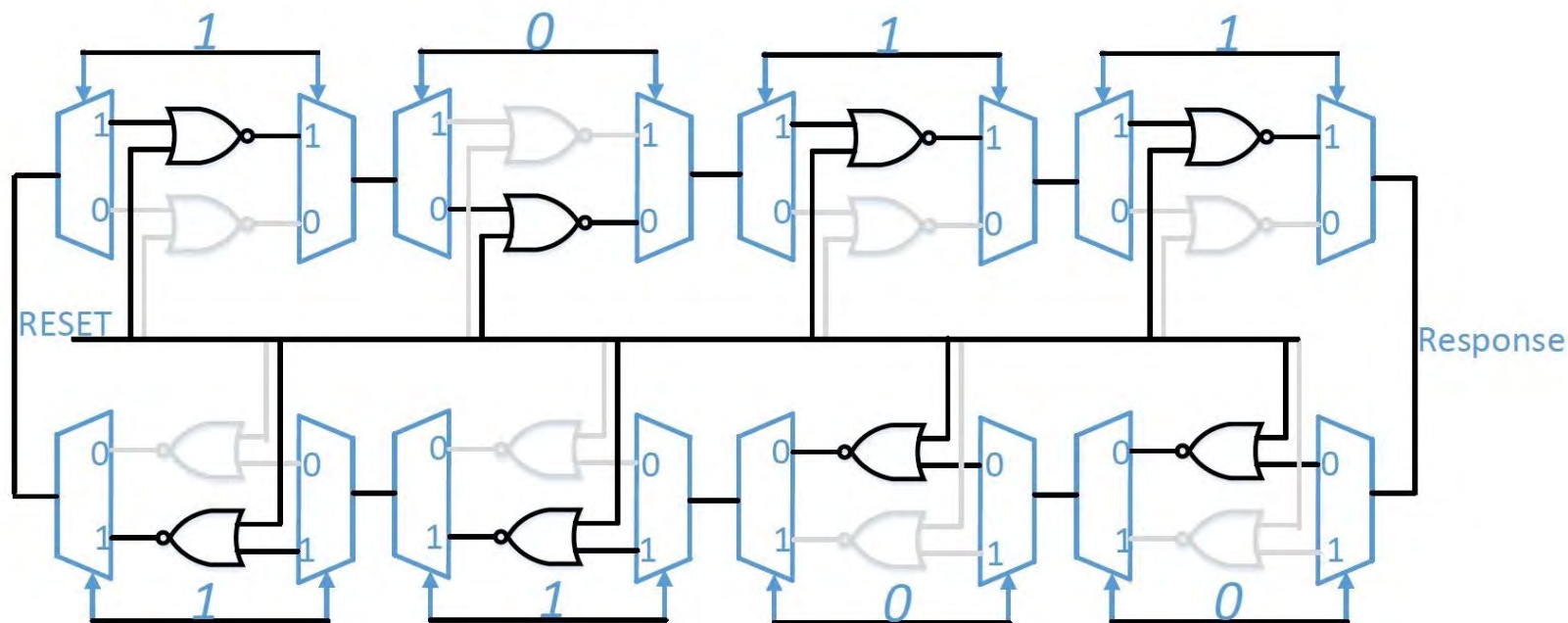
$$t_0 - b_1 + t_2 - t_3 + b_4 - b_5 + t_6 - t_7$$



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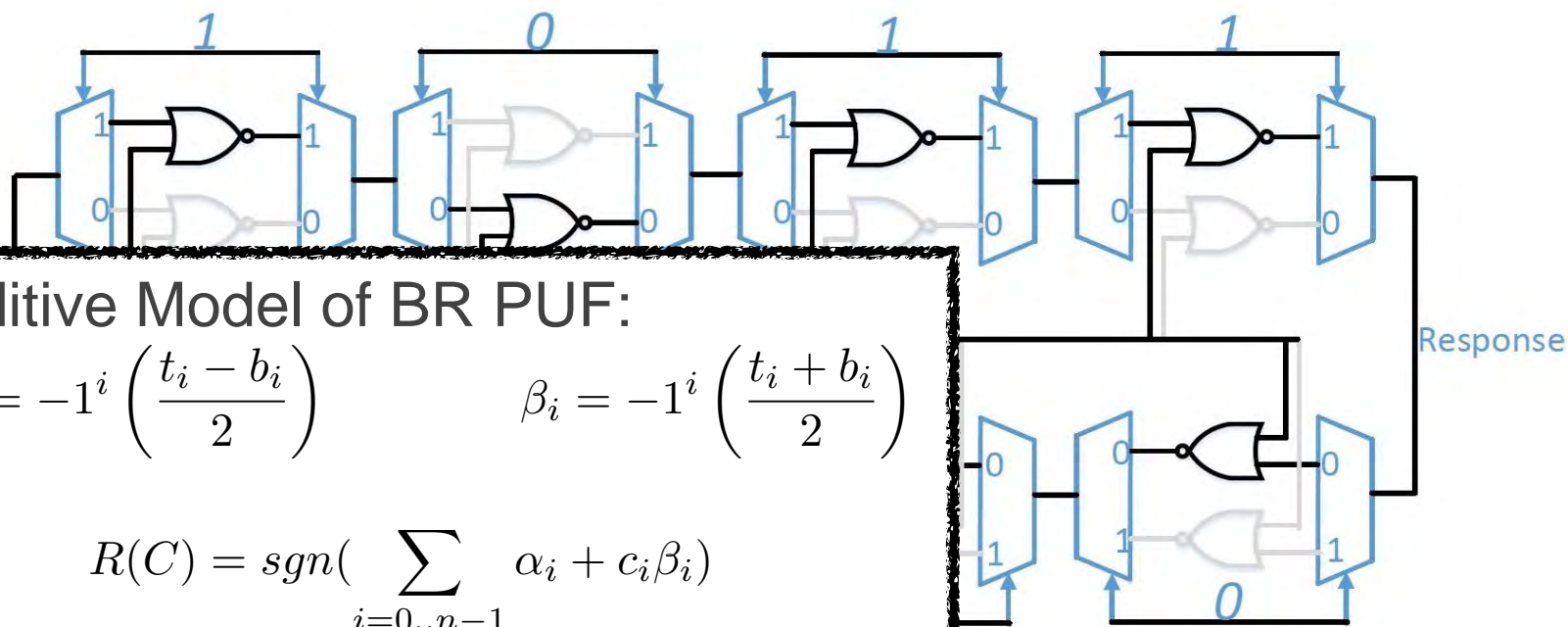
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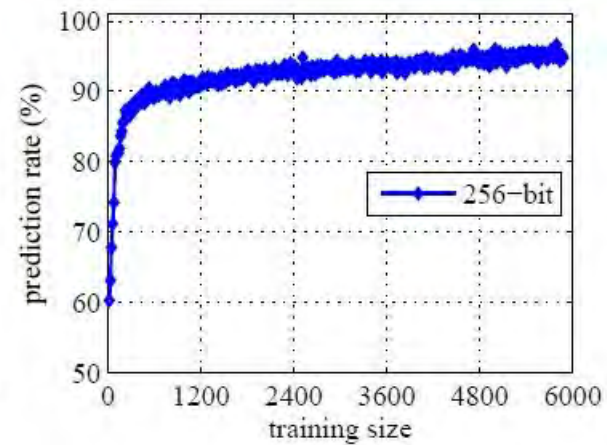
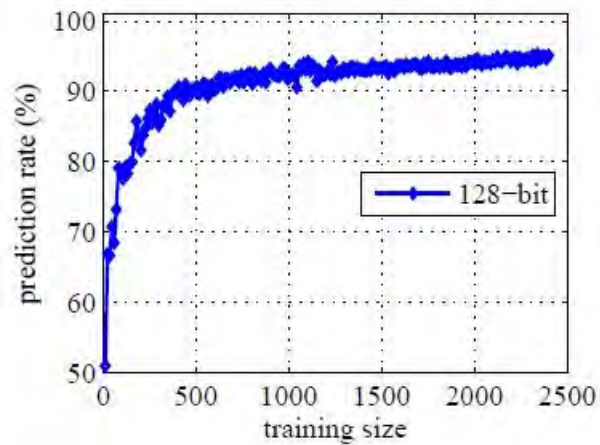
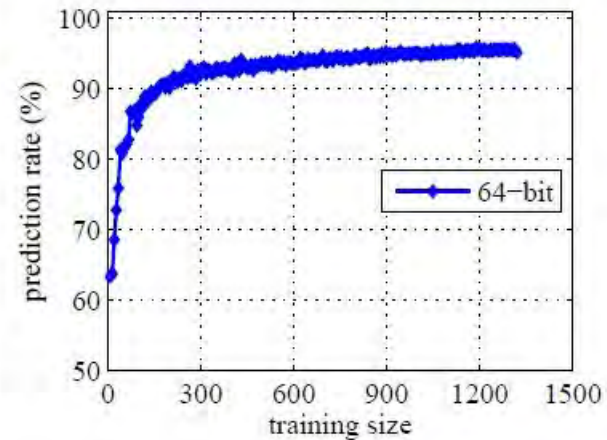
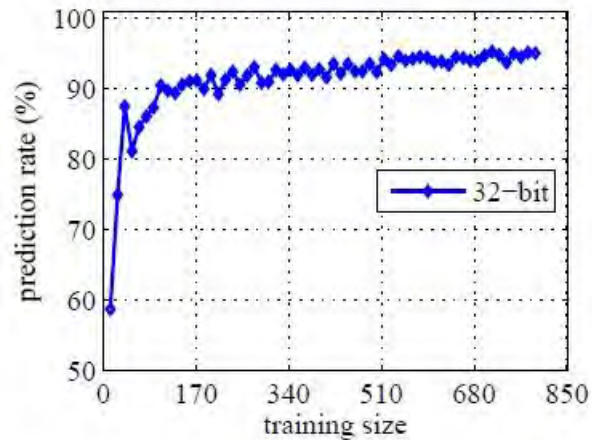


see also Schuster et al. Trust 2014

Implementation of SVM Modeling Attacks

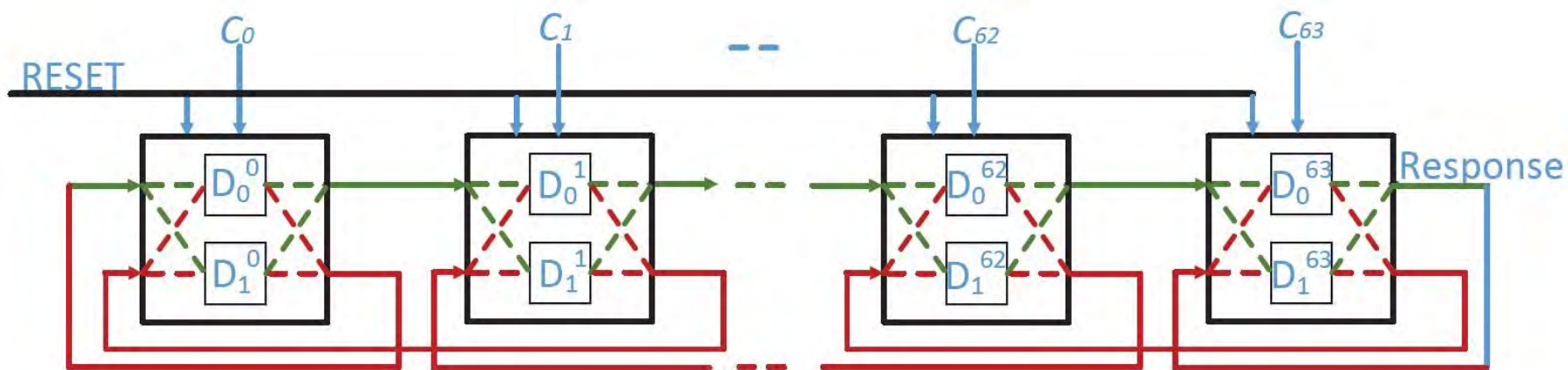
- Modeling with Support Vector Machines classification
- CRPs from FPGA implementation, SVM attacks use Matlab
- CRPs divided into training and validation datasets:
- **Train** the PUF model:
PUF_model=svmtrain(training_input, training_output, 'options',
'kernel_function', 'polynomial', 'polyorder', number_of_XOR); %%
polynomial kernel is used, while the polyorder is the XOR complexity, i.e.,
for a single BR PUF, number_of_XOR=1
- **Validate** the PUF model:
model_output= svmclassify(PUF_model,validation_input);
prediction_rate=(model_output==validation_output); %% predication rate
is the percentage of model_output equals with that of validation output

BR PUF is Not Secure



Twisted BR PUF

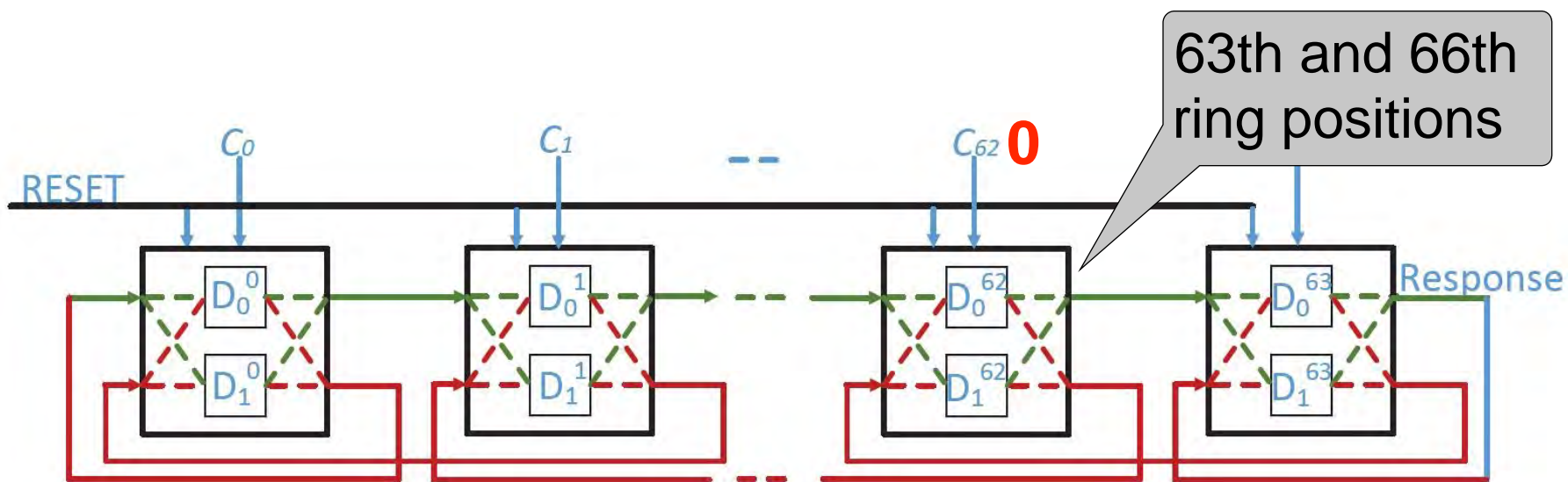
- TBR-PUF⁽⁶⁾ has a more compact design
- All $2n$ inverting elements used in each ring
- Challenge bit determines whether ring position of each inverting element is even or odd
- Additive model still applies and is simpler than regular BR PUF



(6) D. Schuster, et al. *Trust and Trustworthy Computing* 2014

Twisted BR PUF

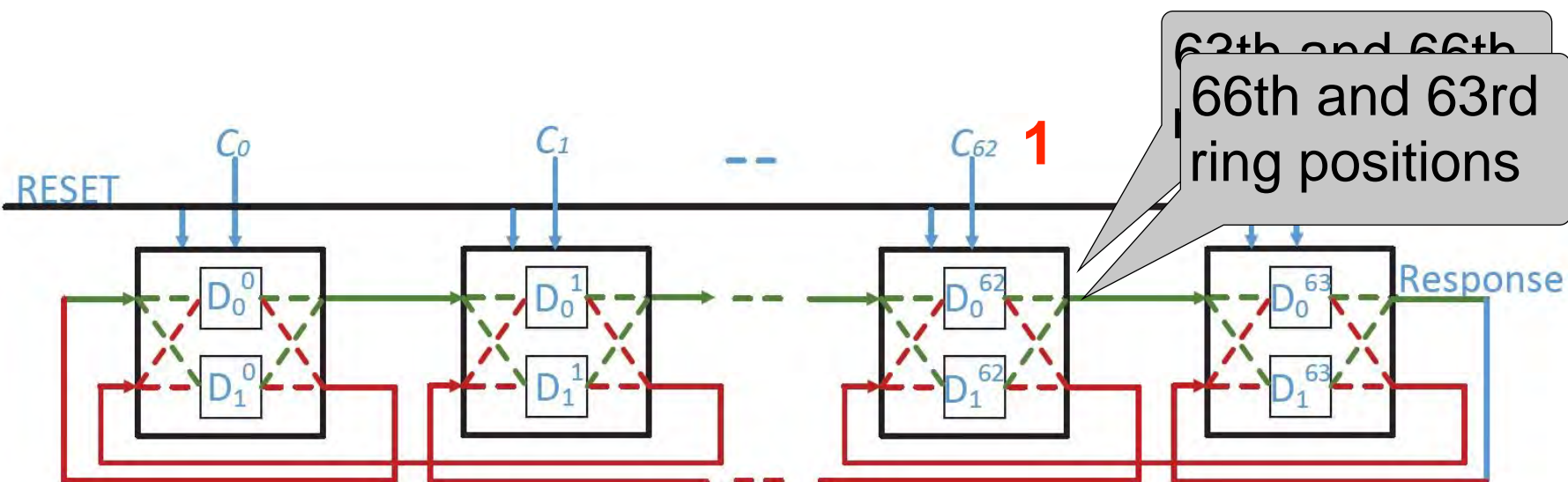
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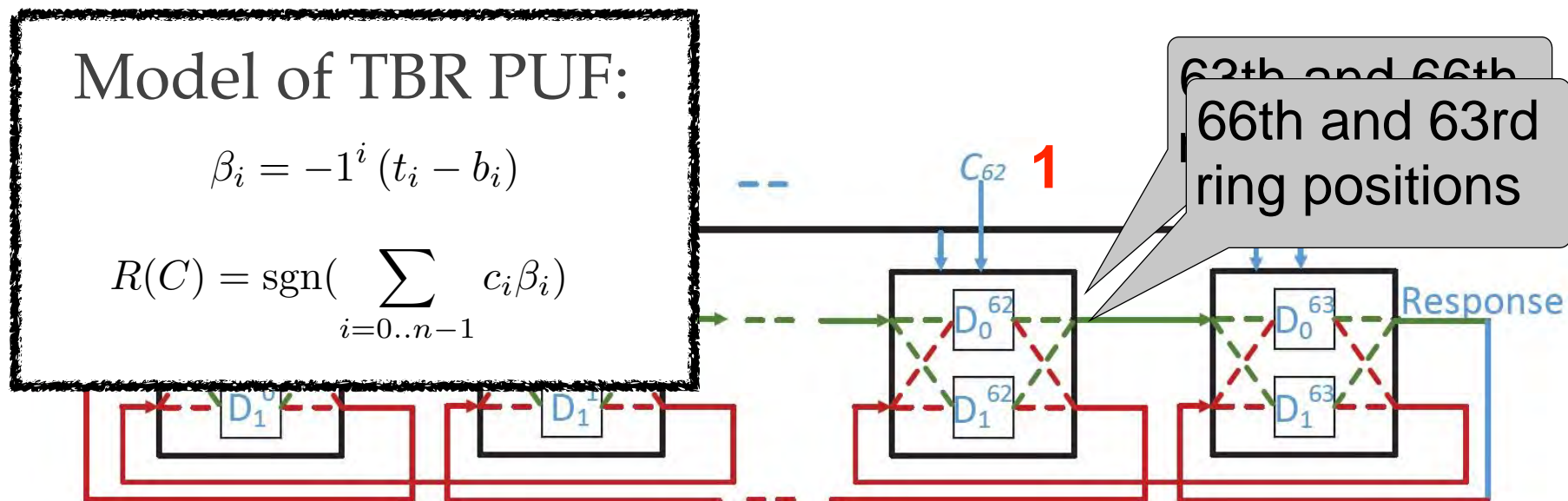
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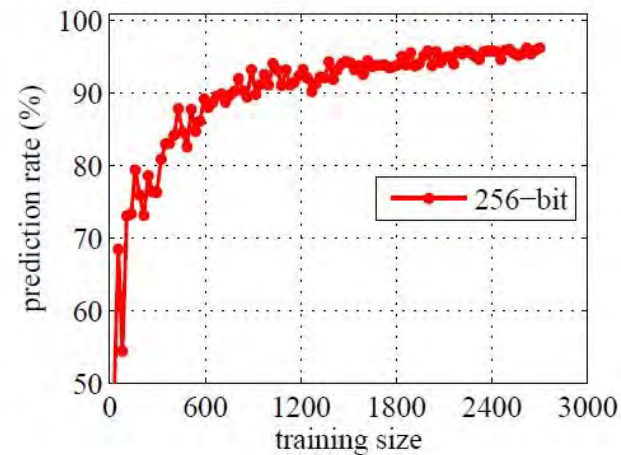
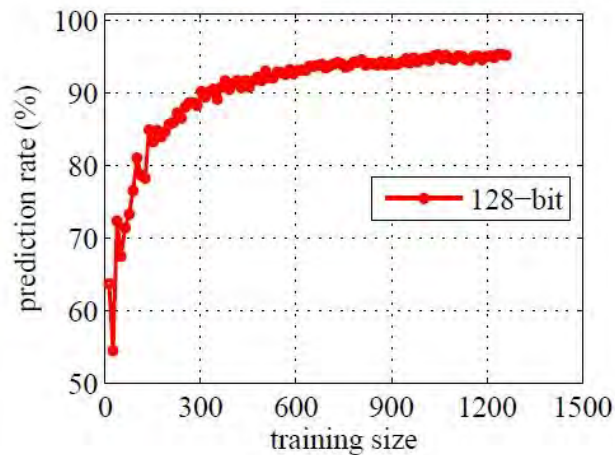
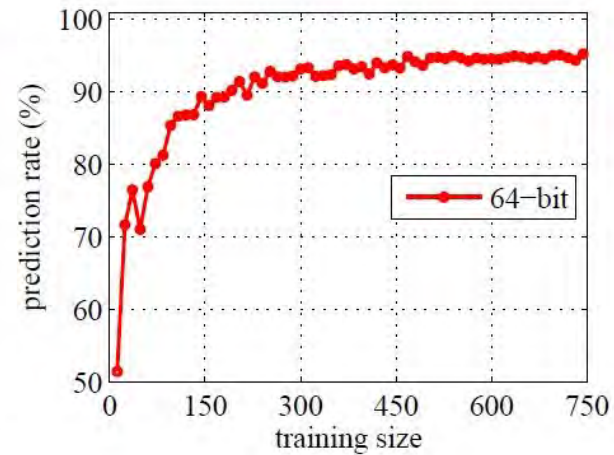
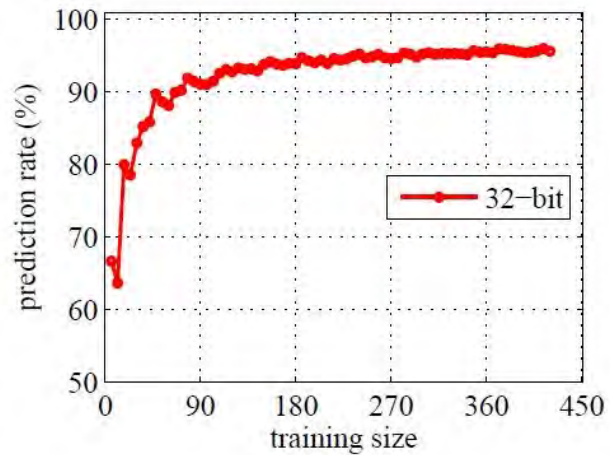
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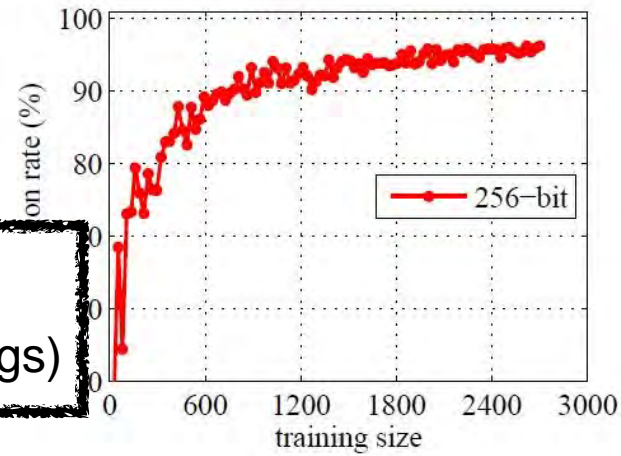
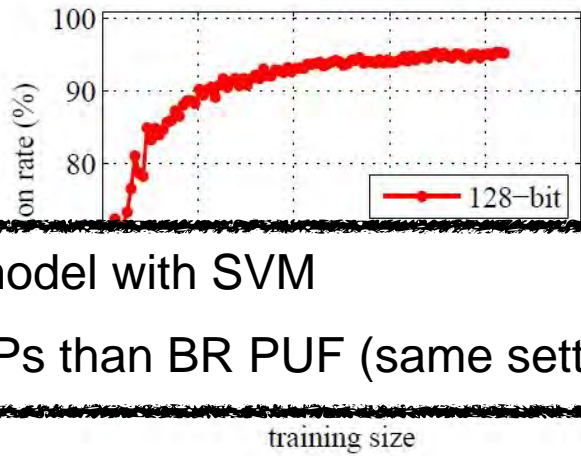
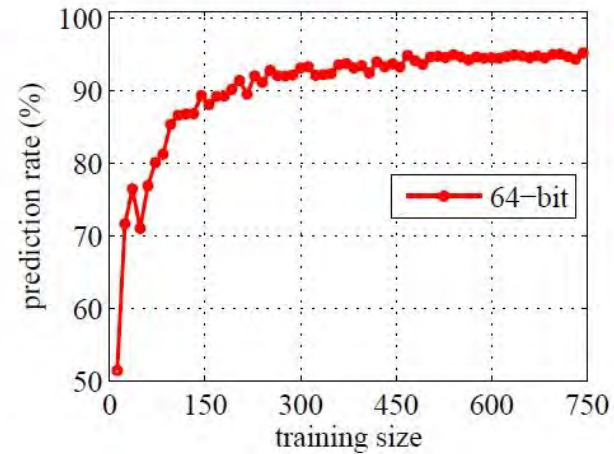
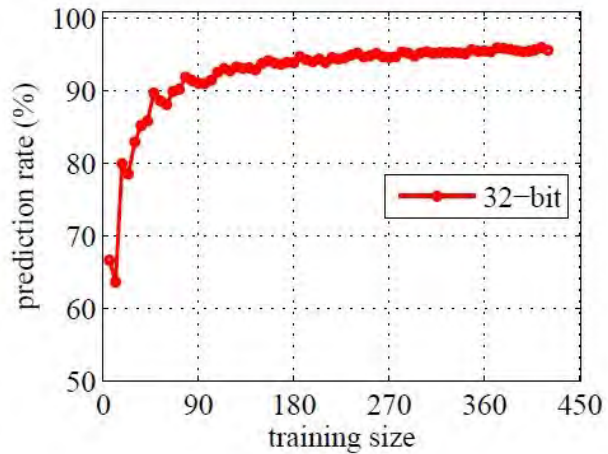


(6) D. Schuster, et al. *Trust and Trustworthy Computing* 2014

TBR PUF is Not Secure



TBR PUF is Not Secure



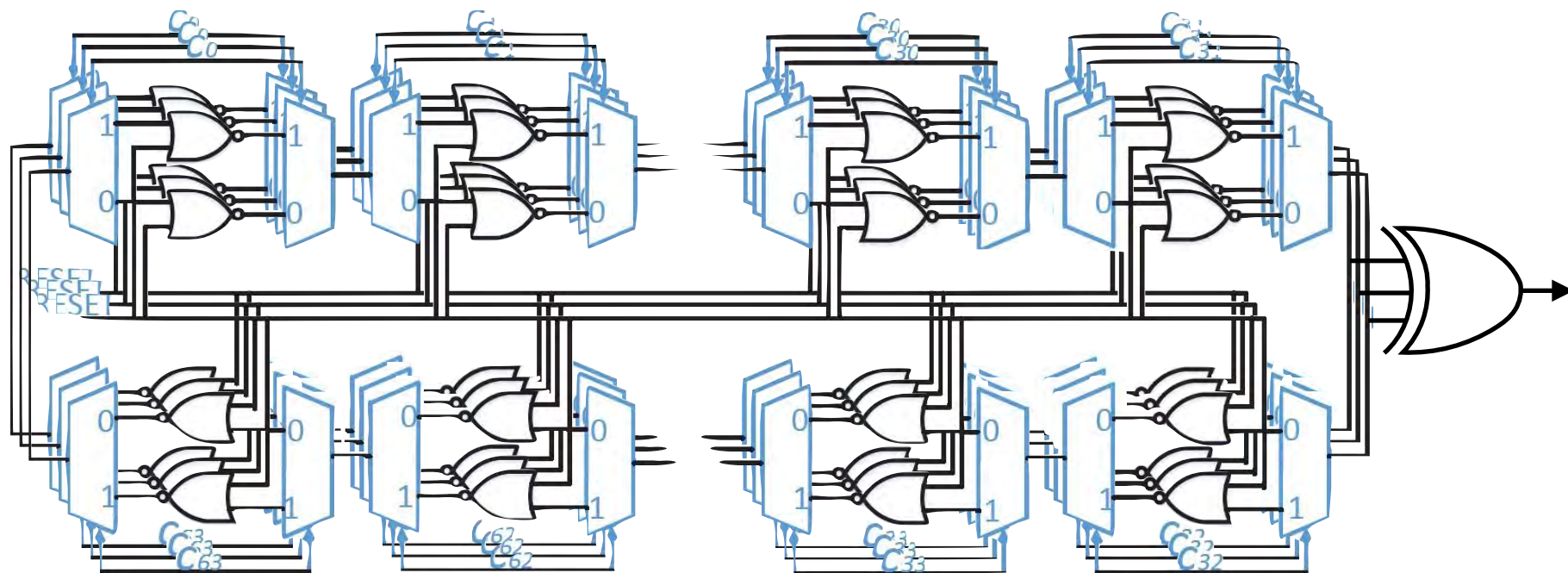
- Trivial to model with SVM
- Fewer CRPs than BR PUF (same settings)

Outline

- Background
 - PUFs
 - Modeling attacks on PUFs
 - Bistable Ring PUF
- Security Evaluation of BR PUFs
 - Modeling the BR PUF
 - Results against BR PUF and variants
- **Security Enhancement of BR PUFs**
 - XORing BR PUFs to enhance the security
 - Impact on other PUF parameters
- Conclusion and future work

XOR BR PUFs to Enhance Security

- XOR responses to harden against SVM modeling attacks
- Prevent direct observation of CRP relation of single PUFs
- Standard technique in many PUF protocols



Security of XOR BR PUFs

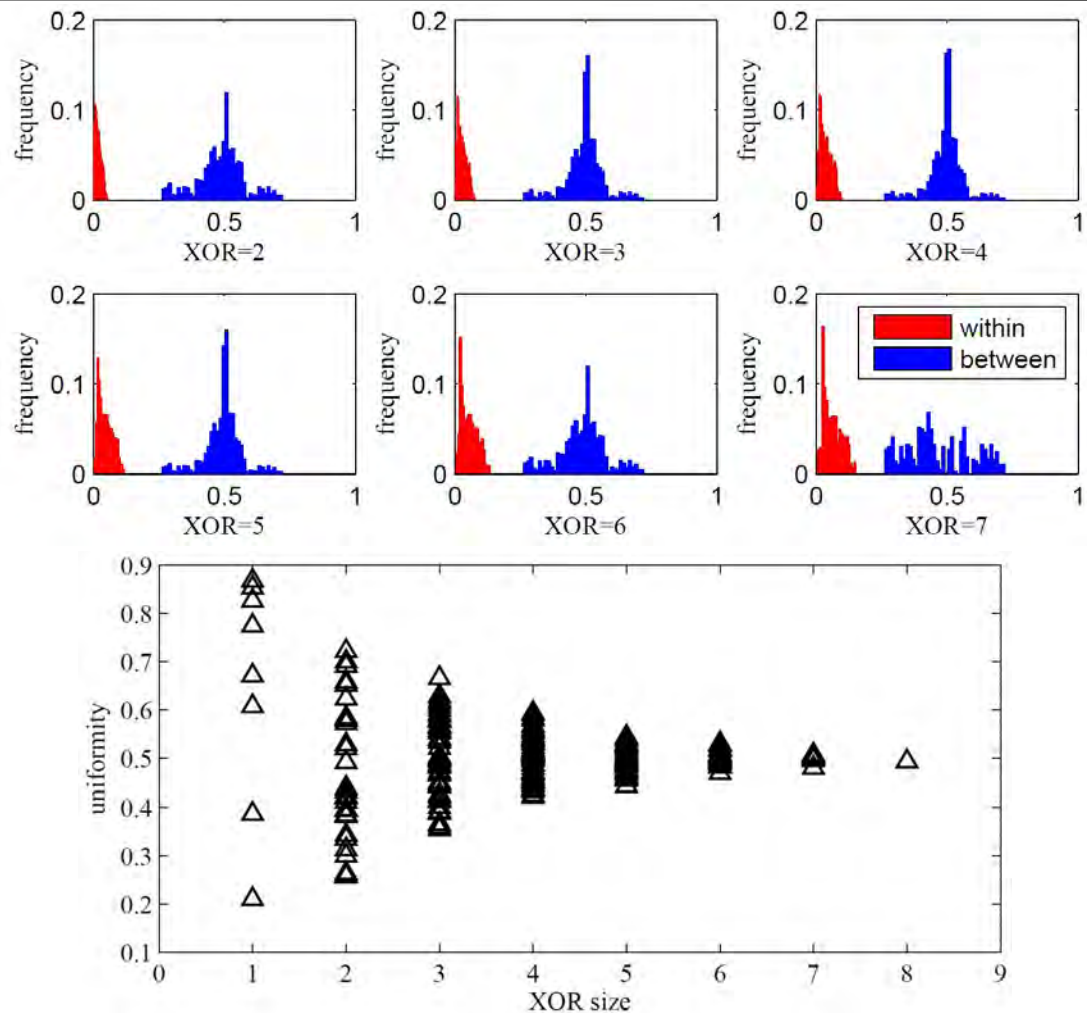
- Resists SVM modeling attacks when >4 XORs used
 - Similar to findings with Arbiter PUFs⁽¹⁾
- Polynomial kernel; polynomial order set equal the number of XORs
- Stronger machine learning attacks may succeed

No. of XORs	Bit Length	CRPs ($\times 10^3$)	Predict. Rate	Training* Time
2	32	0.8	95%	3 sec
	64	4	95%	10 sec
	128	18	95%	6 mins
	256	—	50.8%	—
3	32	1.2	95%	5 sec
	64	7.2	95%	24 sec
	128	—	50.1%	—
	256	—	50.1%	—
4	32	—	50.1%	—
	64	—	50.3%	—
	128	—	49.8%	—
	256	—	50.1%	—

(1) U. Rühmair, et al, CCS, 2010.

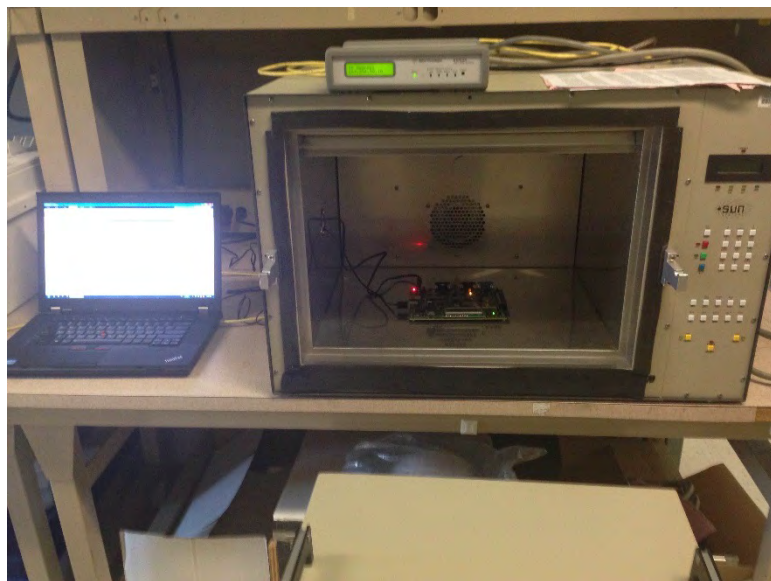
Impact of XOR on Uniqueness and Uniformity

- XOR increases within-class Hamming Distance
- Within-class and between-class HD remain separable
- Single PUFs have poor uniformity
- Uniformity improves with XOR

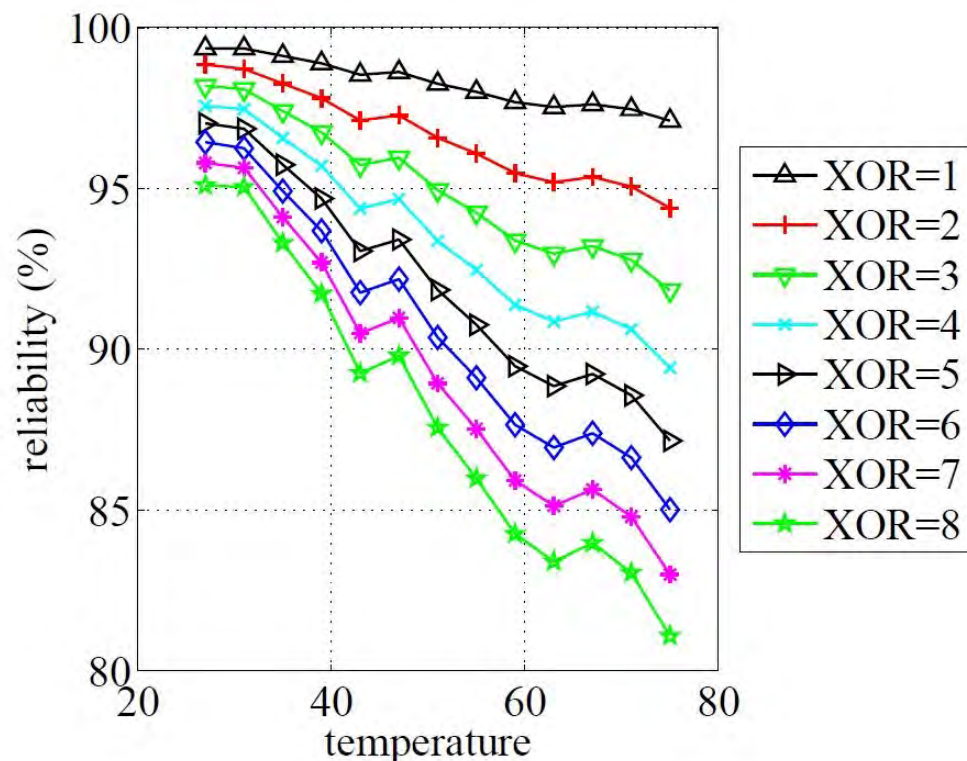


Impact of XOR on Reliability

- XOR degrades reliability
- Any single PUF response flip will change response parity



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Conclusion and Future Work

- BR PUF and TBR PUF are vulnerable to machine learning modeling attacks
 - 95% accurate prediction surpasses capabilities of ANN-based attacks⁽¹⁾
 - Reasonable runtime and fewer than 10k CRPs
- XORing four or more BR PUFs produces a behavior that is beyond the modeling capability of the applied SVM attacks
 - XOR function improves uniformity but degrades reliability
- Future work will explore the effectiveness of other modeling attacks including evolutionary strategies and logistic regression

Thank you for your attention

(1) Schuster et al. TRUST 2014