

High Precision Discrete Gaussian Sampling on FPGAs

Selected Areas in Cryptography 2013

Sujoy Sinha Roy, Frederik Vercauteren and Ingrid Verbauwhede

ESAT/COSIC and iMinds, KU Leuven

Outline of Talk

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- Introduction
- Implementation of discrete Gaussian sampling using Knuth-Yao Random Walk
 - Basics of Knuth-Yao sampling
 - Implementation of Knuth-Yao random walk using counters
 - Space optimization for Probabilities
 - Hardware architecture
 - Results

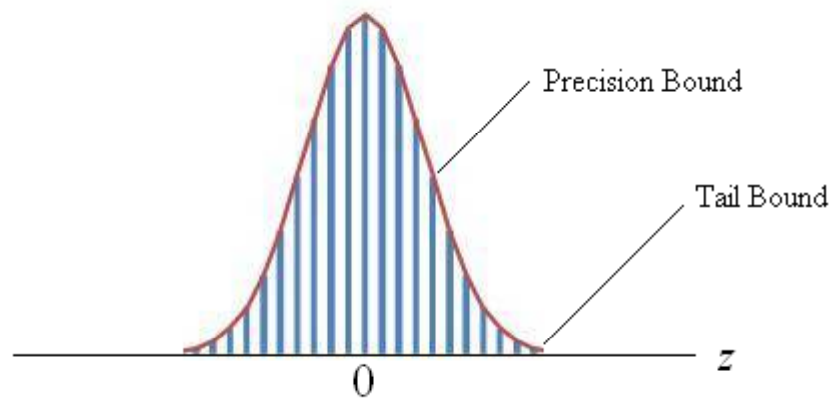
Discrete Gaussian Sampling

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- Discrete Gaussian distribution $D_{\mathbb{Z},\sigma}$ over \mathbb{Z} with mean 0 and standard deviation σ

$$\Pr(E = z) = \frac{1}{S} e^{-z^2/2\sigma^2} \quad \text{where } S = 1 + 2 \sum_{z=1}^{\infty} e^{-z^2/2\sigma^2}$$

- Tail is infinitely long
- Probabilities have infinite precision

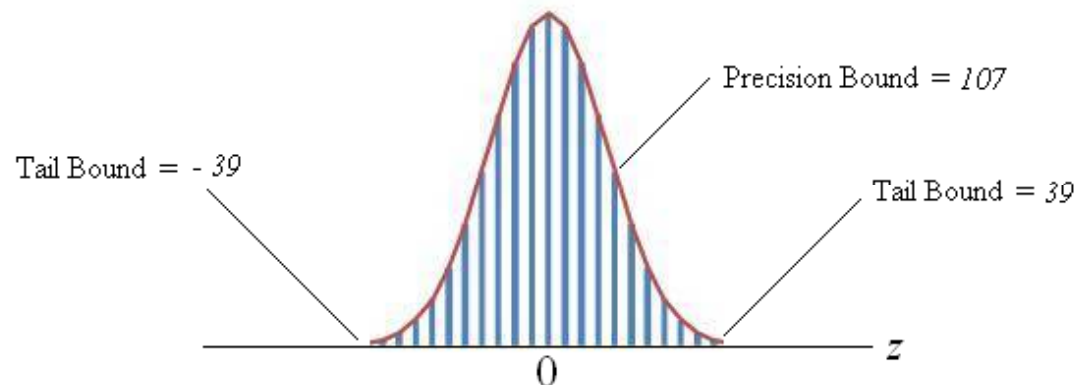


Discrete Gaussian Sampling : Tail/Precision Bounds

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- Provable Security :
 - Negligible statistical distance from true Gaussian distribution : 2^{-90}
 - For standard LWE parameter set
 - Tail bound : practically 39

m	s	σ	Tail	Precision
256	8.35	3.33	84	106
320	8.00	3.192	86	106
512	8.01	3.195	101	107



Sampling Methods

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- Commonly used methods
 - Rejection sampling
 - Inversion sampling
- Large number of random bits are required to maintain high precision
- Slow on resource-constrained platforms

Knuth-Yao Sampling

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- Random-walk model
- Requires near-optimal number of random bits
- Example : Let a sample space $S = \{0, 1, 2\}$

$$p_0 = 0.01110$$

$$p_1 = 0.01101$$

$$p_2 = 0.00101$$

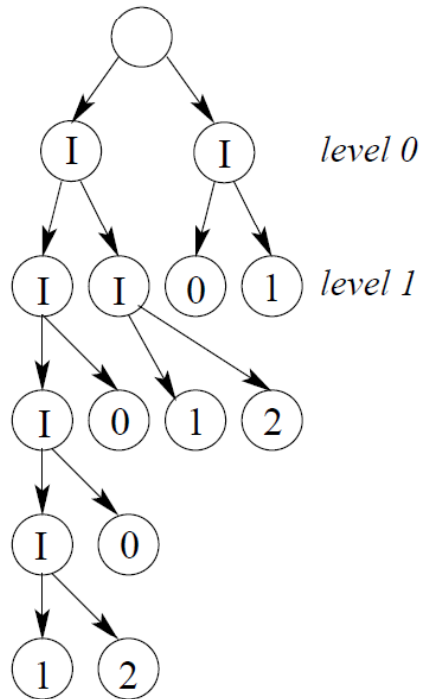
- Probability matrix

$$P_{mat} = \begin{matrix} \begin{matrix} \text{row } 0 \rightarrow \\ \text{row } 1 \rightarrow \\ \text{row } 2 \rightarrow \end{matrix} & \begin{matrix} \downarrow \text{column } 0 \\ \downarrow \text{column } 1 \\ \downarrow \text{column } 2 \\ \downarrow \text{column } 3 \\ \downarrow \text{column } 4 \end{matrix} & \begin{pmatrix} 0 & 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \end{pmatrix} \end{matrix}$$

Knuth-Yao Sampling

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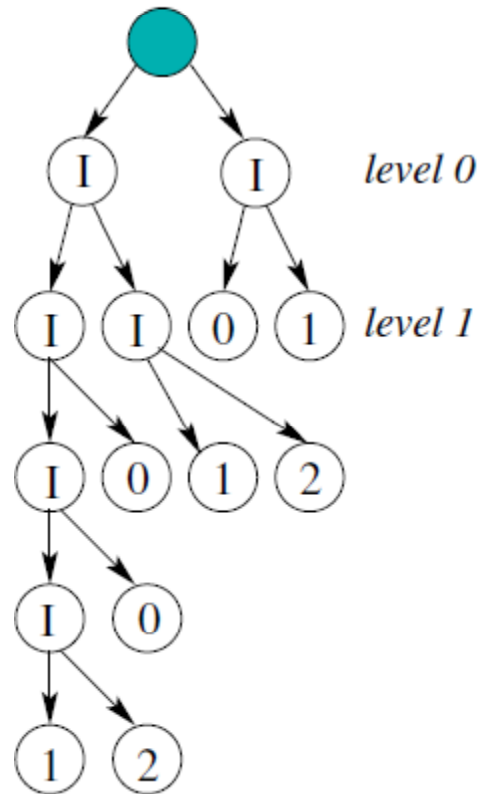
- Discrete Distribution Generating (DDG) tree is formed
 - Binary tree corresponding to P_{mat}



$$P_{mat} = \begin{bmatrix} 0 & 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \end{bmatrix}$$

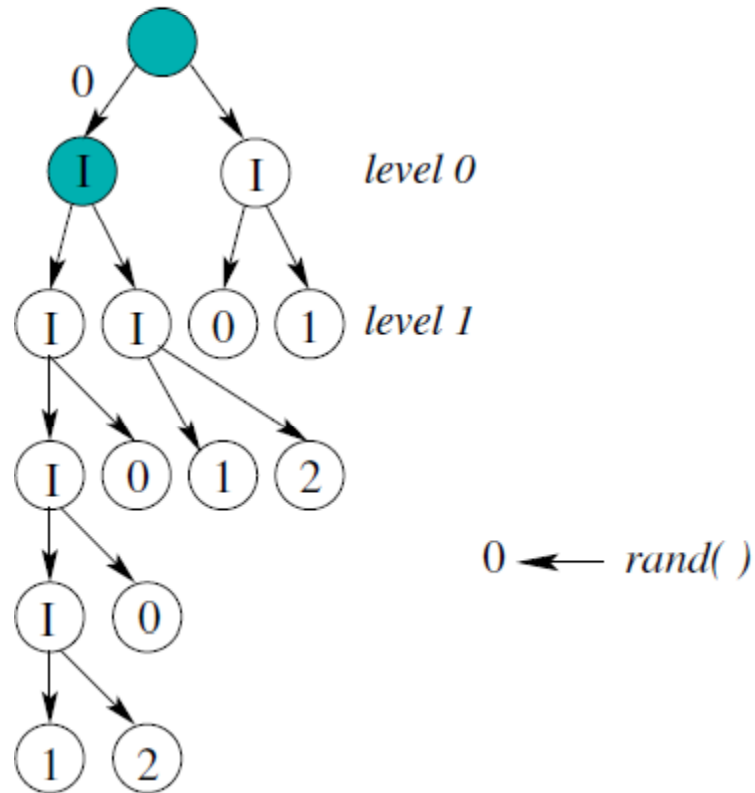
Knuth-Yao Sampling : Random Walk

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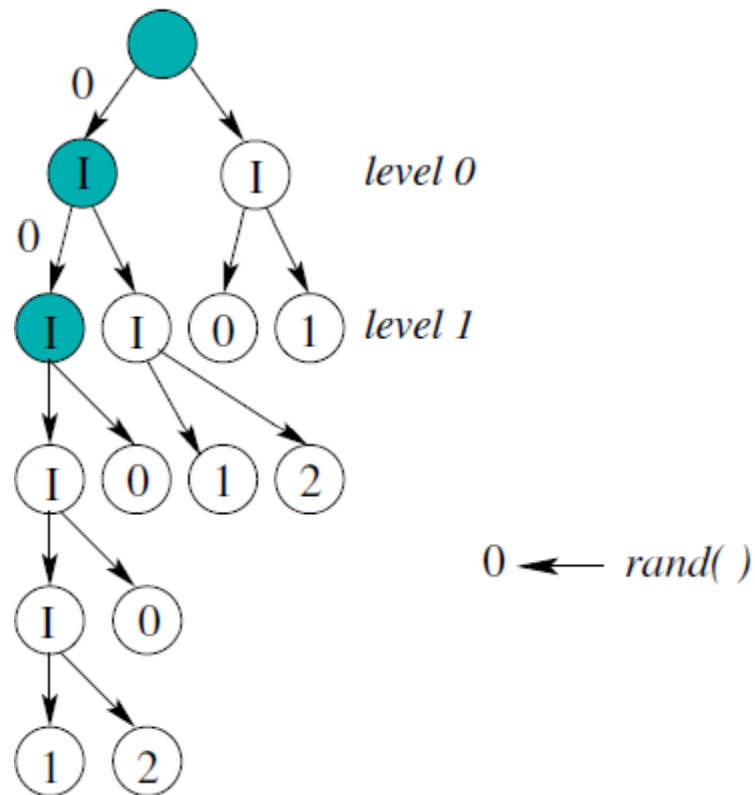
Knuth-Yao Sampling : Random Walk

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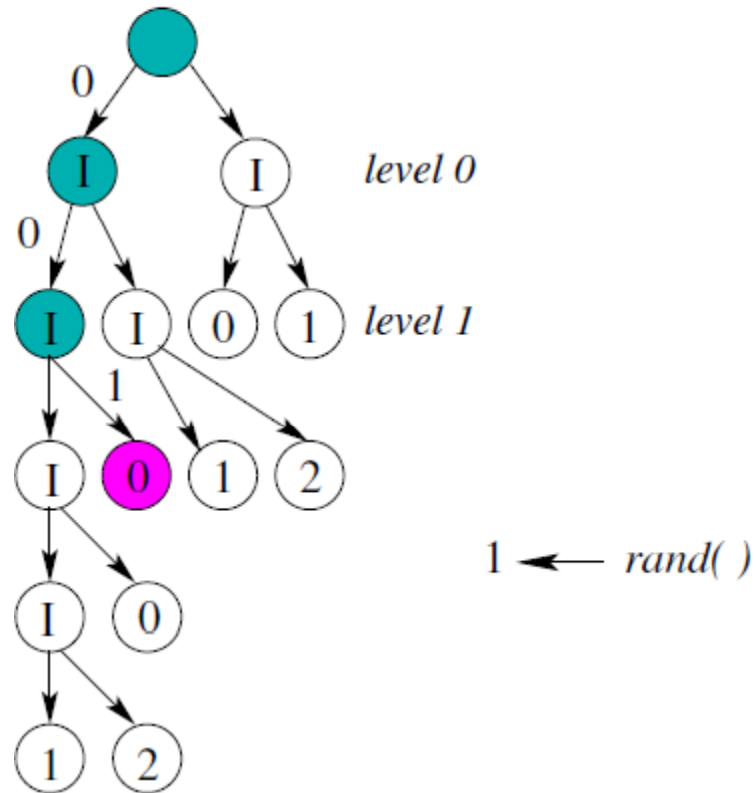
Knuth-Yao Sampling : Random Walk

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Knuth-Yao Sampling : Random Walk

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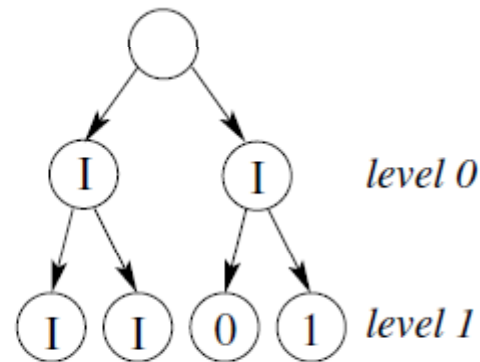


Knuth-Yao Sampling : Implementation

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- Any level of the DDG tree can be constructed from previous level using probability matrix

$$P_{mat} = \begin{pmatrix} 0 & 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 \end{pmatrix}$$

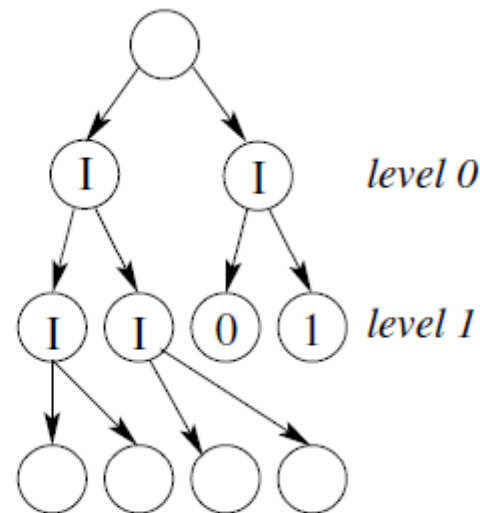


Knuth-Yao Sampling : Implementation

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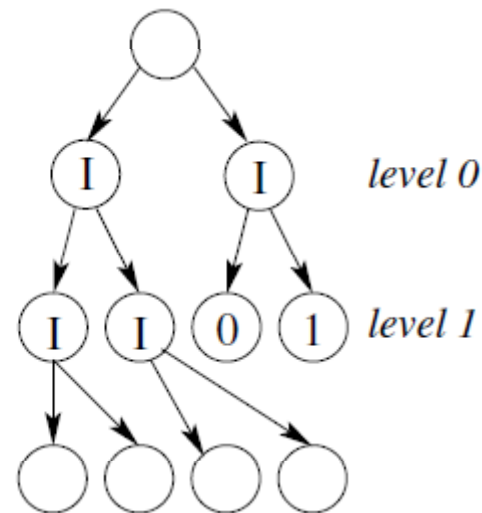


Knuth-Yao Sampling : Implementation

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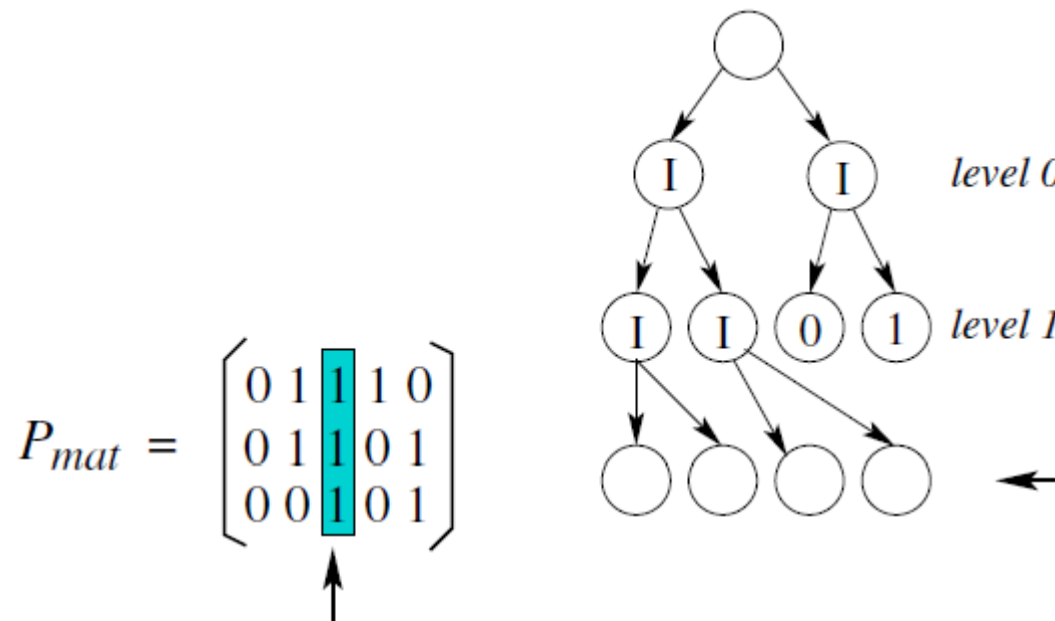
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Knuth-Yao Sampling : Implementation

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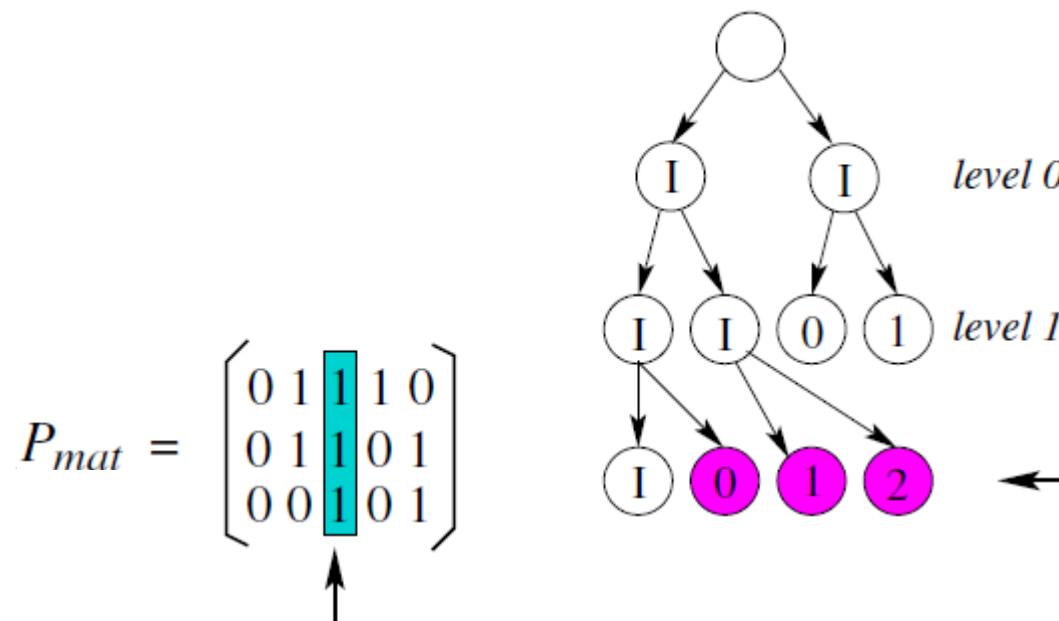
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Knuth-Yao Sampling : Implementation

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- Any level of the DDG tree can be constructed from previous level using probability matrix



Knuth-Yao Sampling : Two Important Points

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- Knuth-Yao random walk
- Storage for Probability Matrix

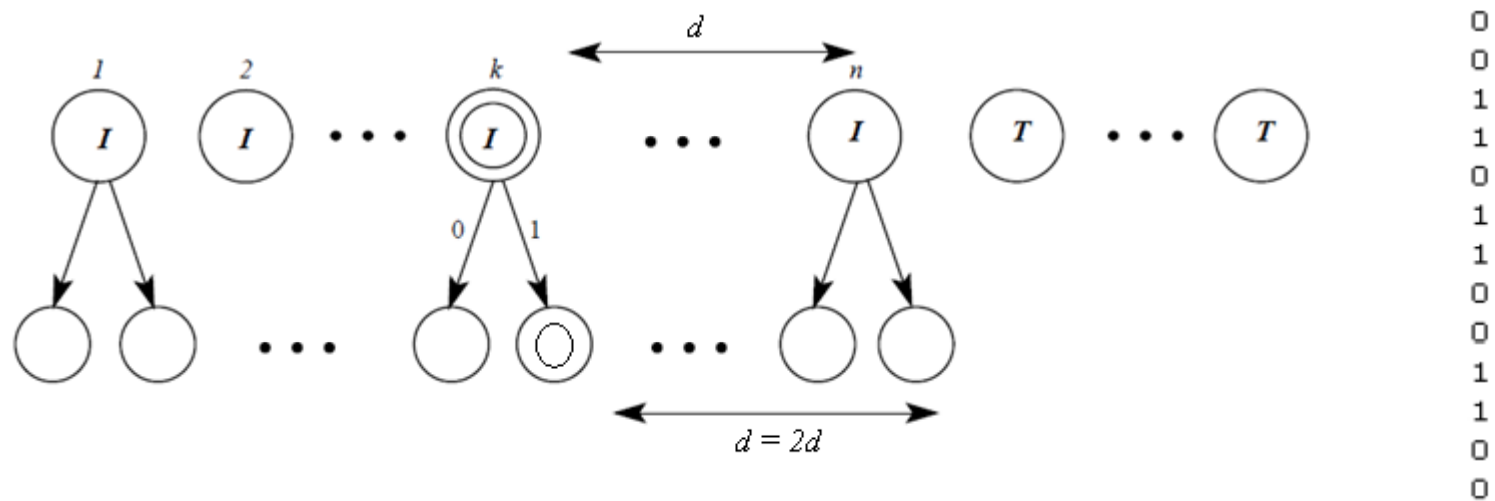
Knuth-Yao Sampling

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Random Walk using Counters

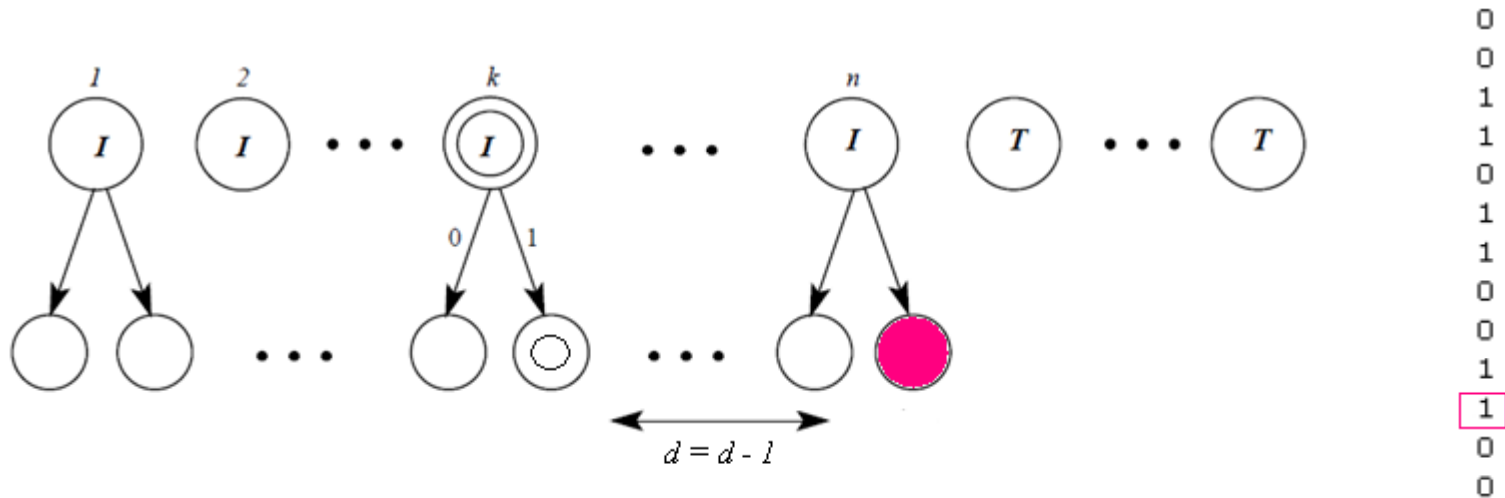
Knuth-Yao Sampling : using Counters

- Construction of i -th level during sampling : Counter d for distance



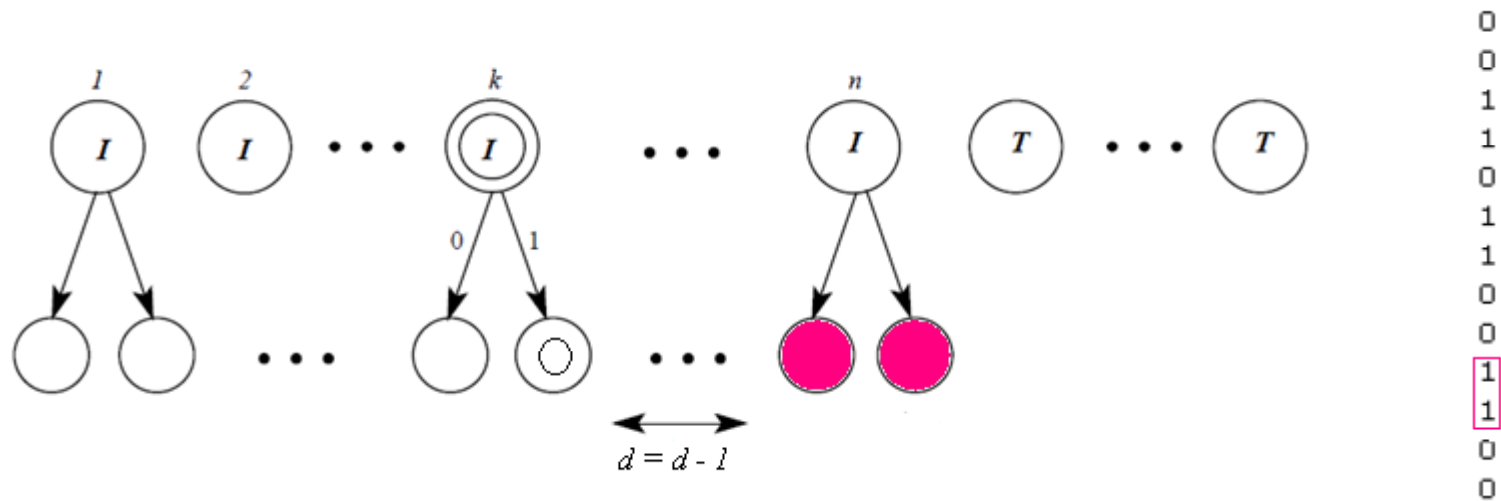
Knuth-Yao Sampling : using Counters

- Construction of i -th level during sampling : Counter d for distance



Knuth-Yao Sampling : using Counters

- Construction of i -th level during sampling : Counter d for distance



- When $d < 0$ for the first time, the visited node is a terminal node
- We need counters for d and **row-number**

Space Optimization for Probability Matrix

Probability Matrix : Column-wise Optimization

- Probability matrix is stored in ROM in a column-wise manner
 - Zeros present in bottom of a column are not stored
 - Lengths of columns are also stored

```
000111111111010111000101110101
001111001101110110011011001101
001101001000110011101100011010
001010010010001110000011001110
000111010011001101100110100000
000100101100101100100011010010
000010101111011110010010001110
000001011100110110001001011000
000000101100100010110010101101
000000010011011000000110100010
000000000111101001000111111011
000000000010101110111011001001
000000000000111000101110001100
000000000000010000101011010101
000000000000000100011100100010
000000000000000001000100110001
00000000000000000001111000100
00000000000000000000010111111
```


Probability Matrix : Column-wise Optimization

- Observation
 - Difference in length is 1 for most consecutive columns

- We consider one-step difference in column length
 - One bit is required per differential column-length
 - 1 for increment
 - 0 for no-increment

```
000111111111010111000101110101
001111001101110110011011001101
001101001000110011101100011010
001010010010001110000011001110
000111010011001101100110100000
000100101100101100100011010010
000010101111011110010010001110
000001011100110110001001011000
000000101100100010110010101101
000000010011011000000110100010
00000000011101001000111111011
000000000010101110111011001001
00000000000011000101110001100
000000000000010000101011010101
00000000000000100011100100010
0000000000000001000100110001
00000000000000001111000100
000000000000000001011111
```

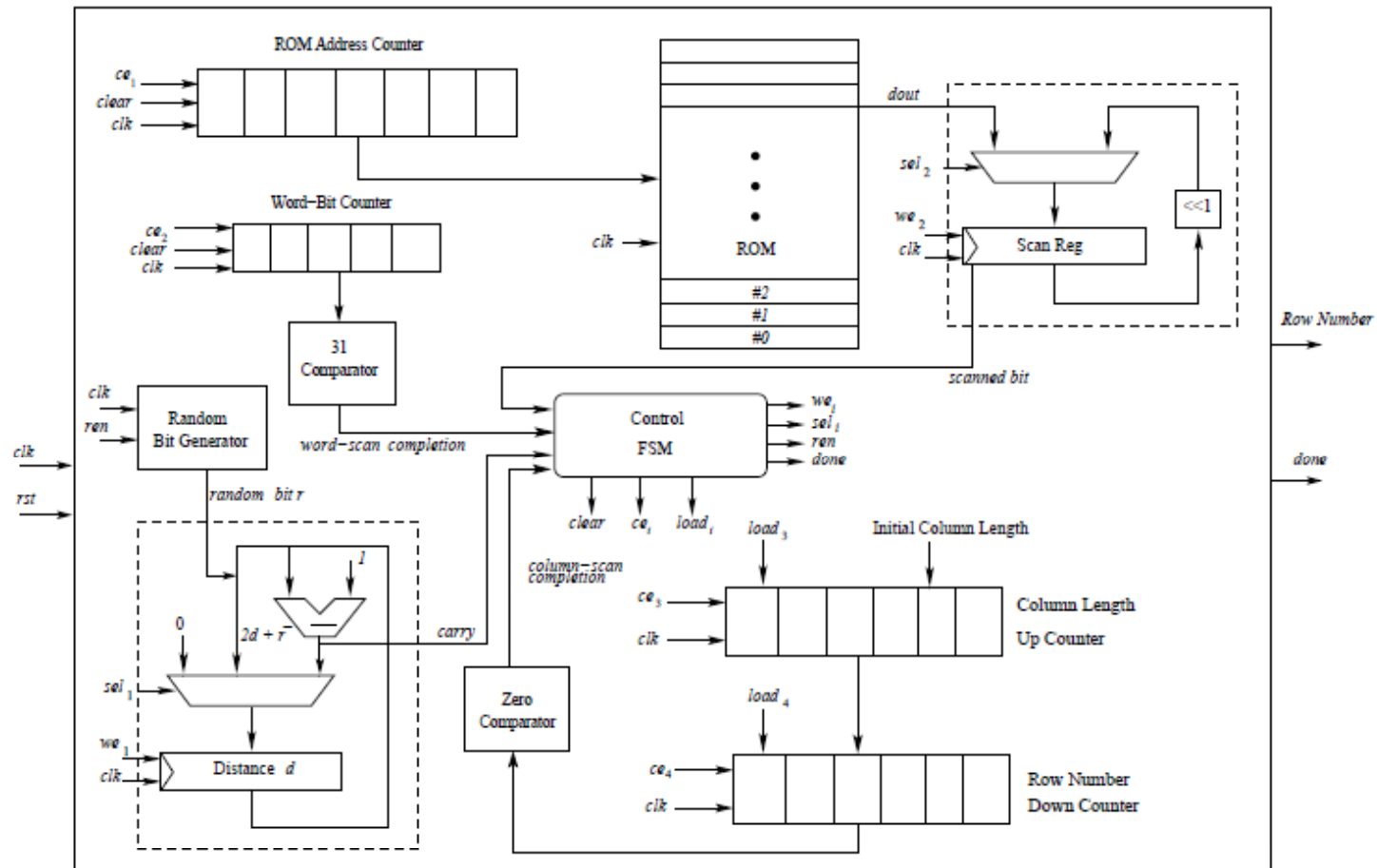
Hardware Architecture

Hardware Architecture

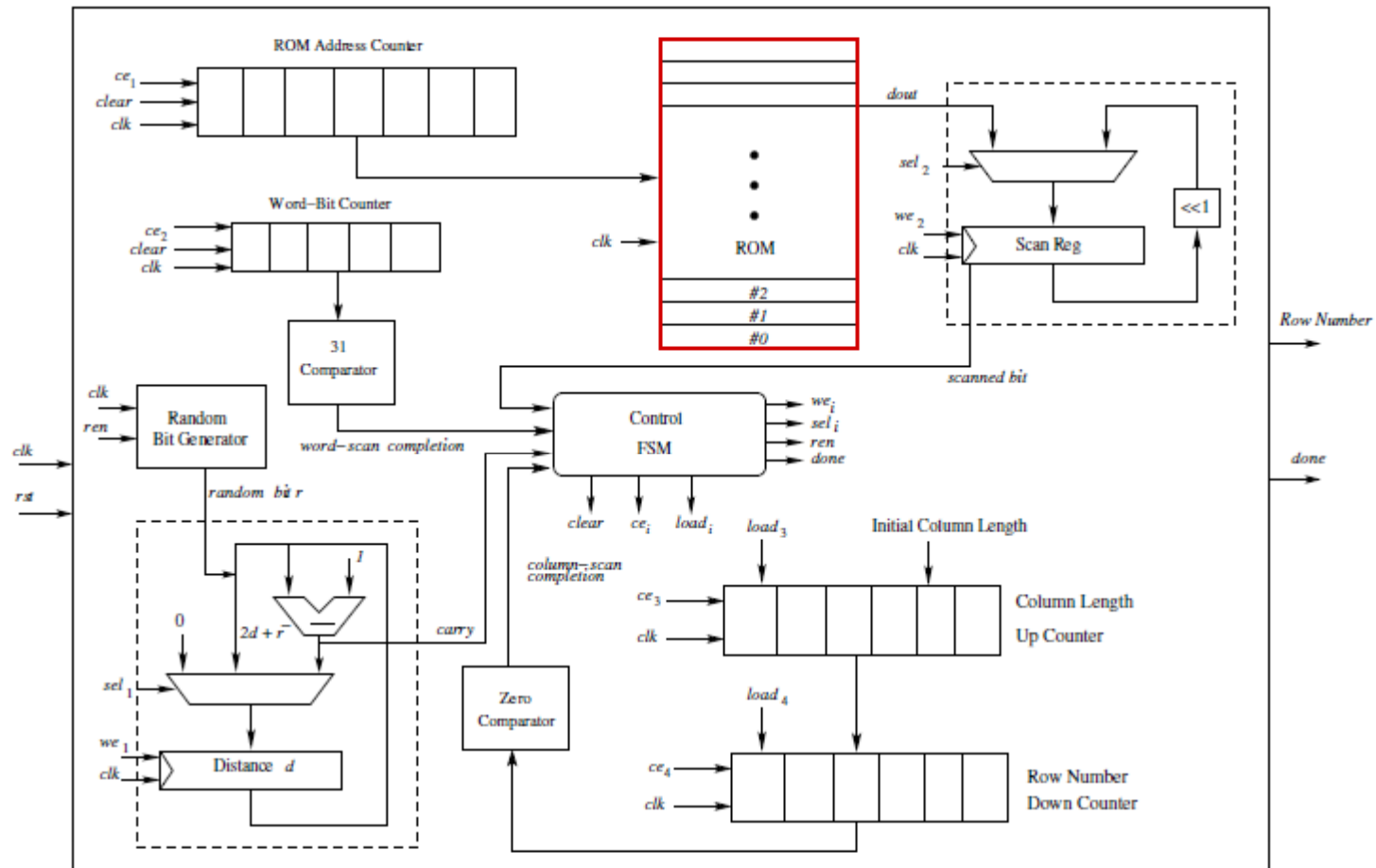
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- Components required
 - ROM for storing probability matrix
 - Counters for d , *row* and *column* during Knuth-Yao sampling
 - Comparators for checking terminal conditions
 - Shift-register for scanning columns

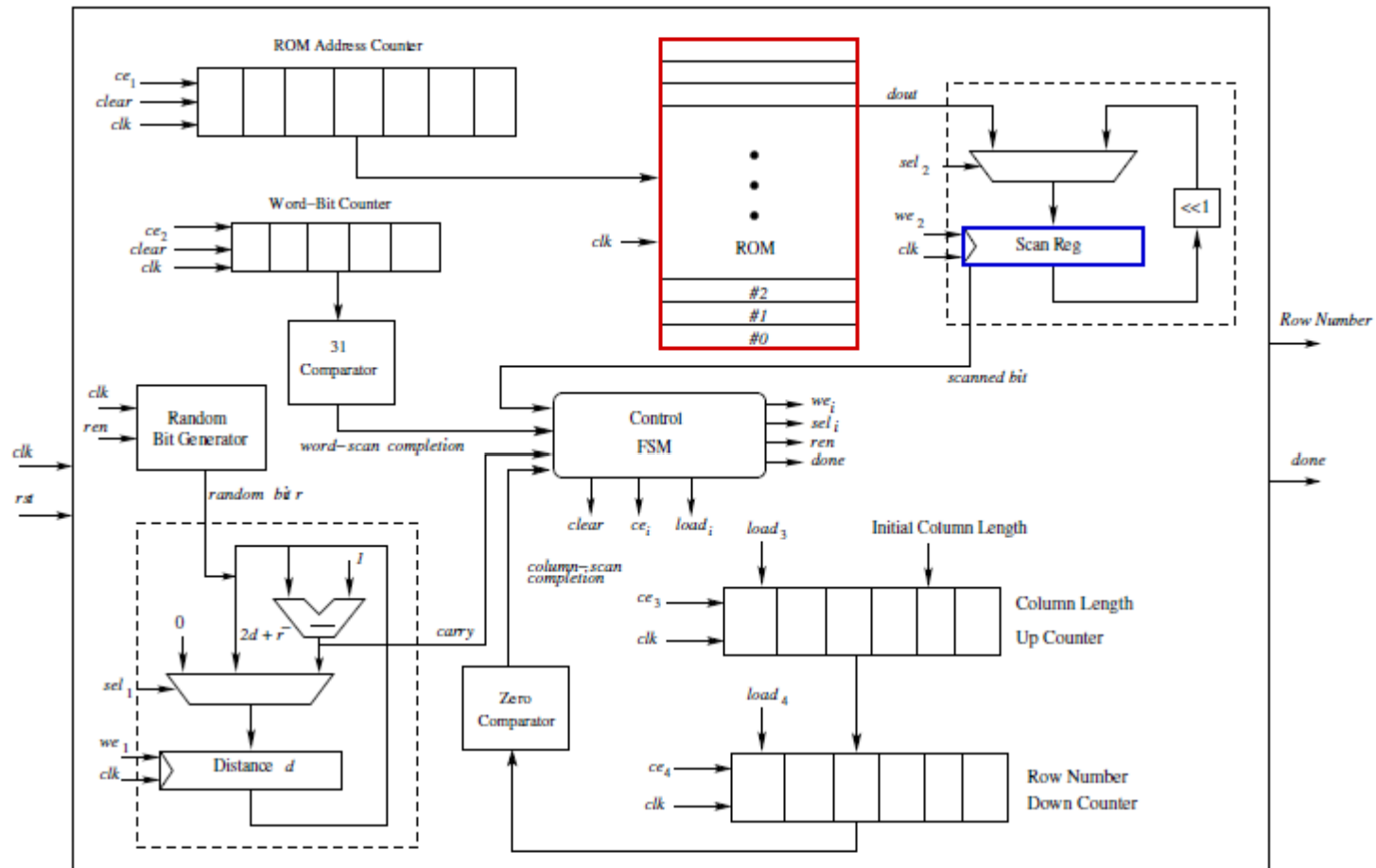
Hardware Architecture



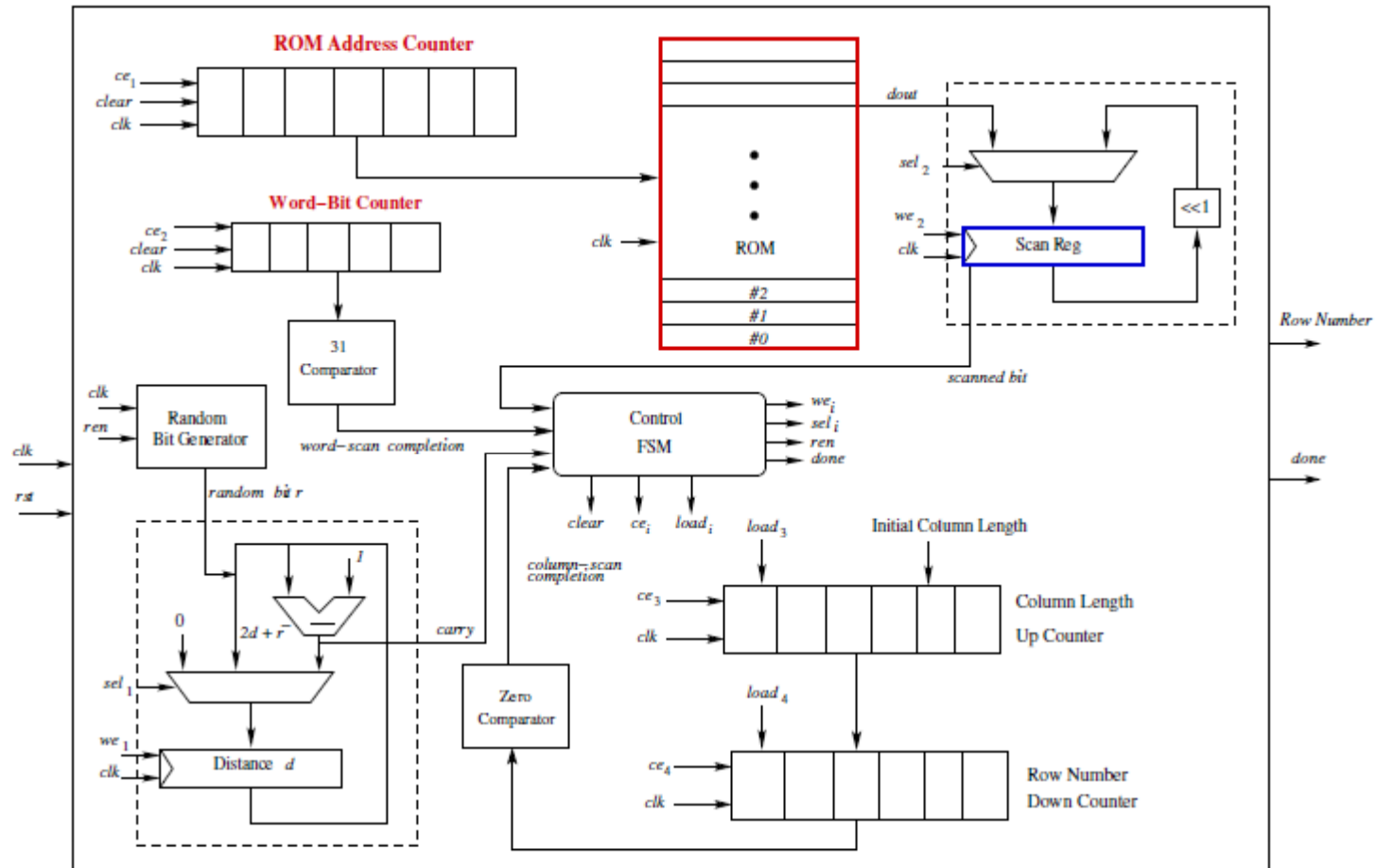
Hardware Architecture



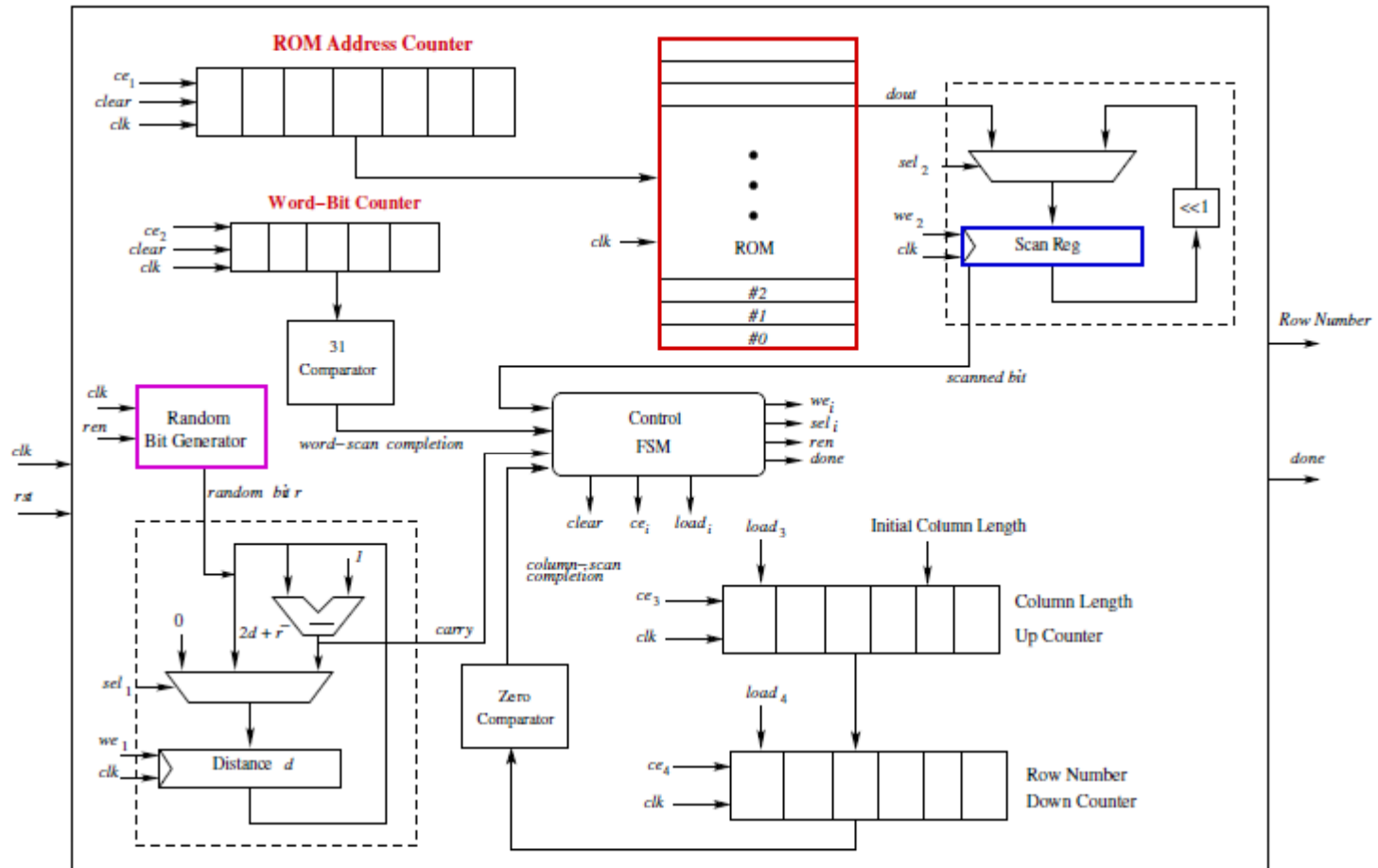
Hardware Architecture



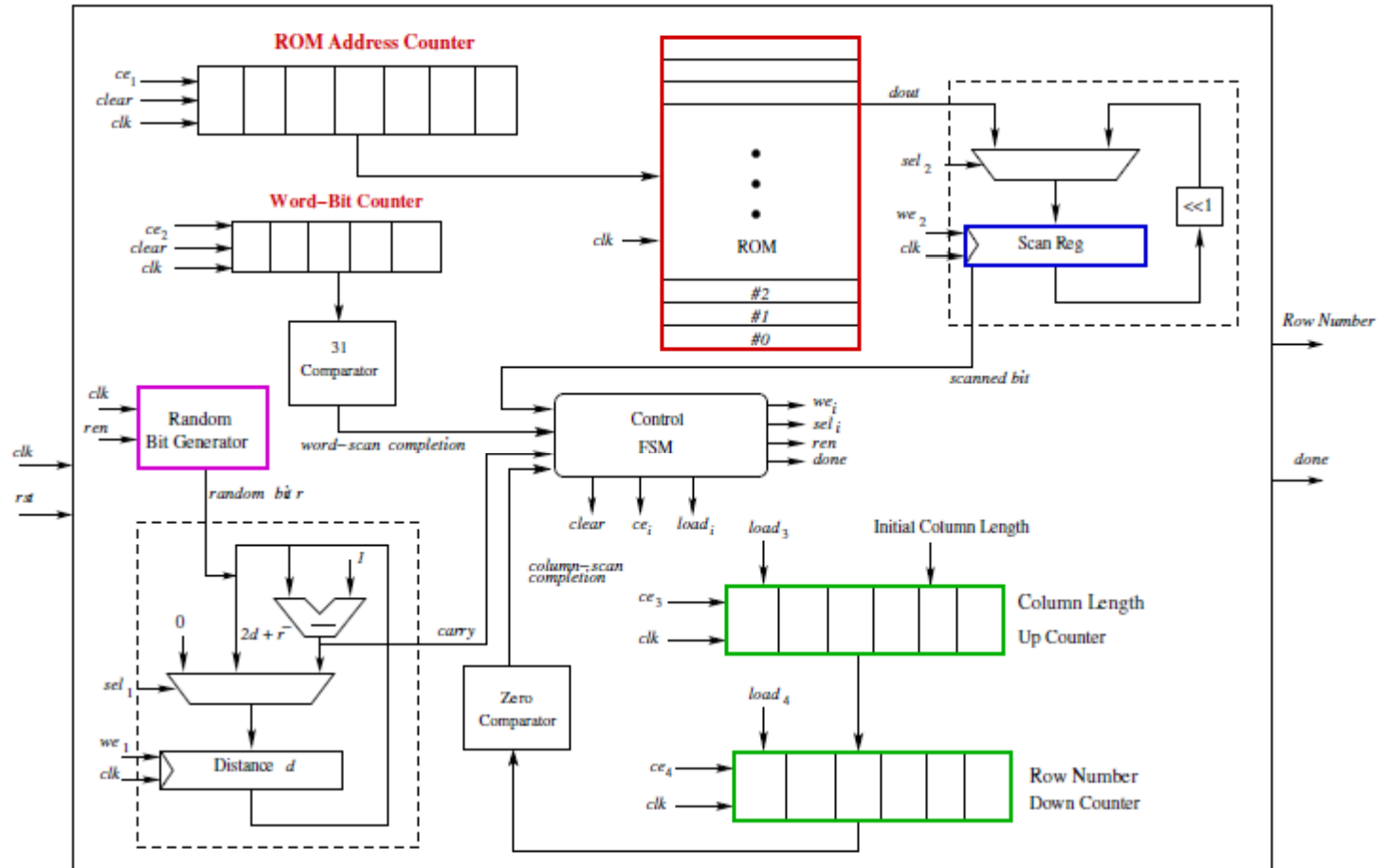
Hardware Architecture



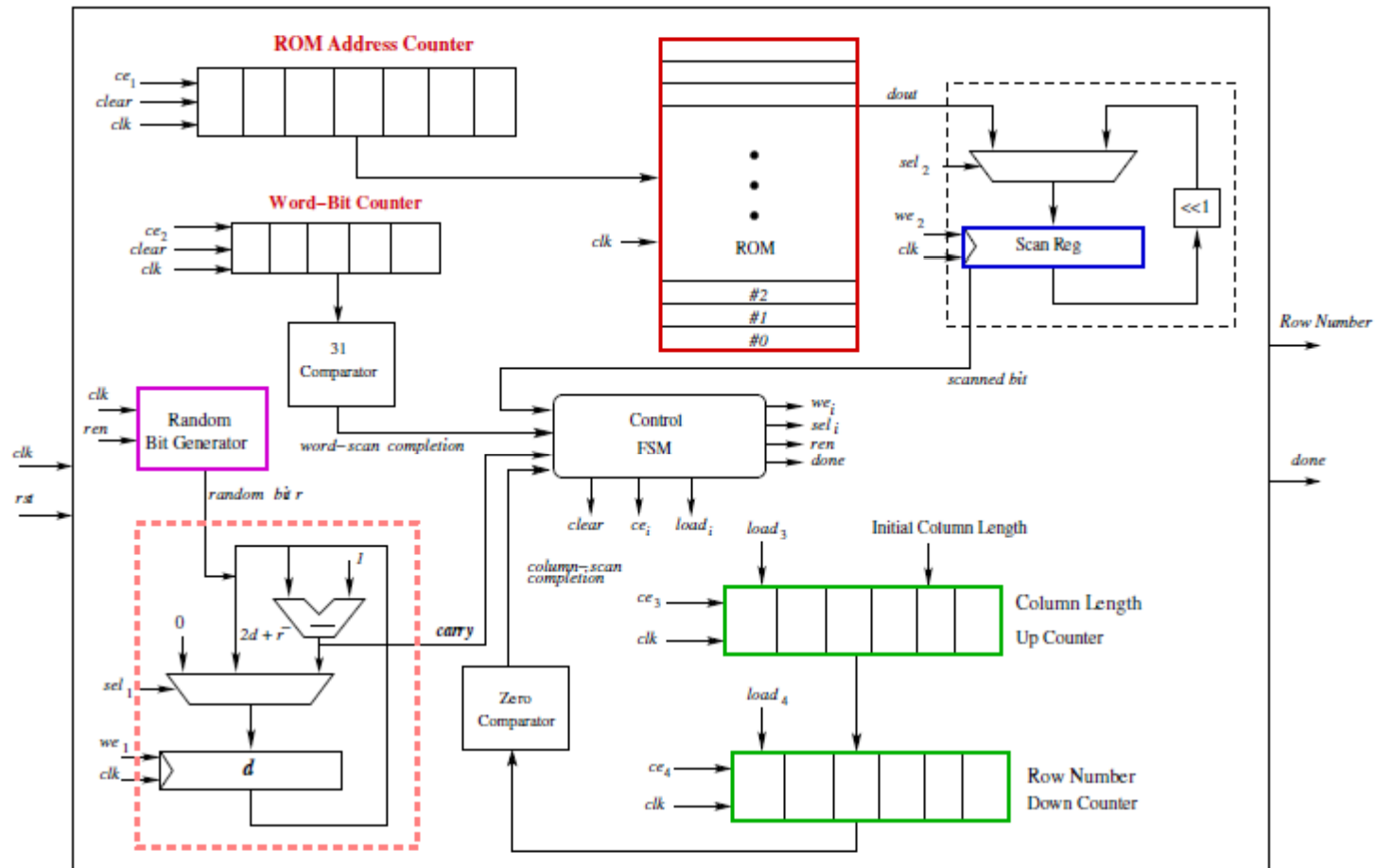
Hardware Architecture



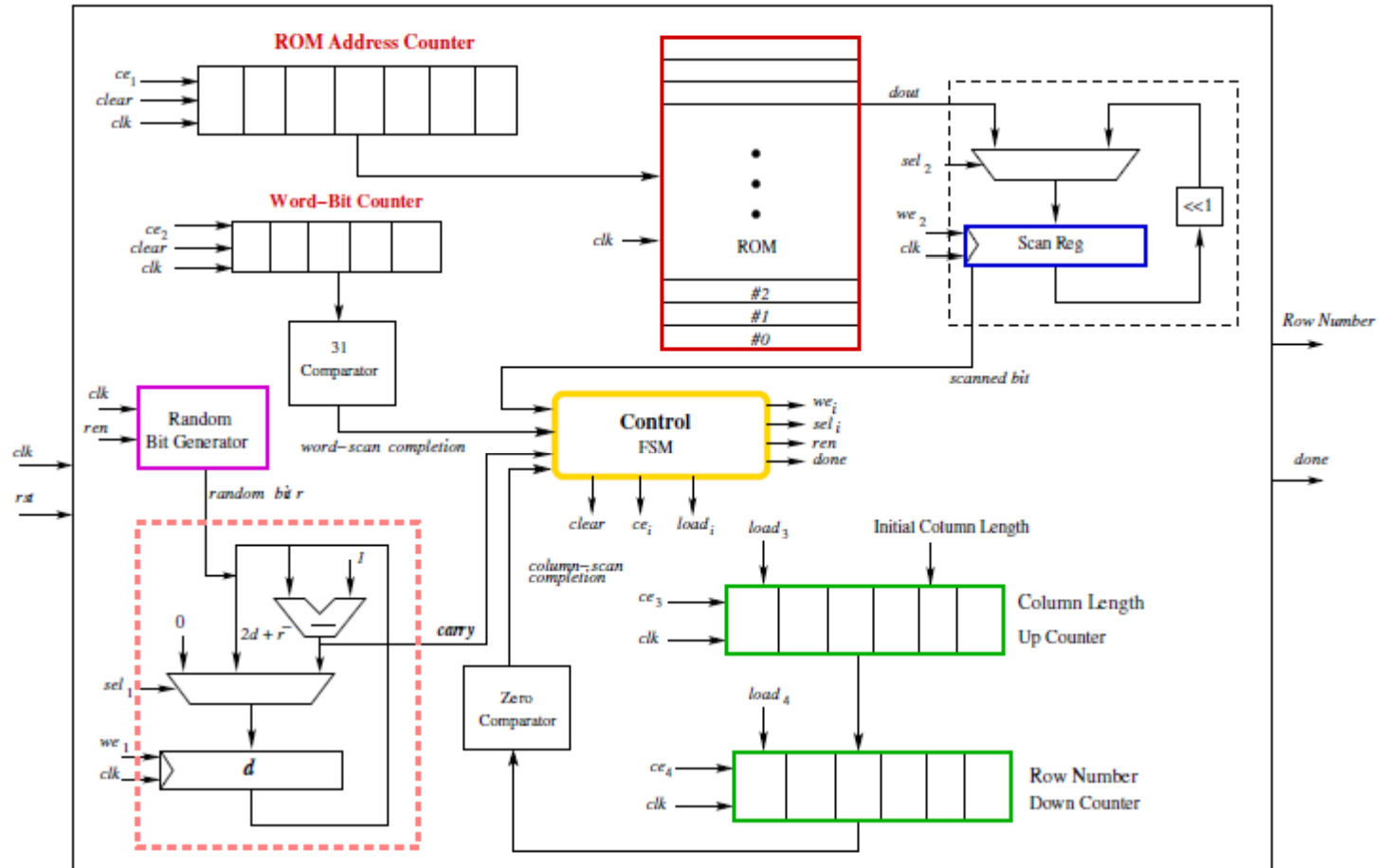
Hardware Architecture



Hardware Architecture



Hardware Architecture



Hardware Architecture : Speeding-Up

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- Scanning of the column bits is the most time consuming operation.
- Hardware => Parallelism
 - Window-based scanning of column bits
 - Reduces scanning time
 - Marginal increase in area

Experimental Results

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Architecture	FFs	Slices		LUTs		Delay (<i>ns</i>)	Clock Cycles
		Core	ROM	Core	ROM		
Basic	66	30	17	76	64	3	17
Window	69	36	17	85	64	3.3	16

Performance of the discrete Gaussian sampler on xc5vlx30

- Storage for Probability matrix
 - 32-by-96 ROM
 - Results do not include the Random Bit Generator
 - Window method provides acceleration for long random walks
 - Timing Analysis is present in the paper

Conclusion & Future Work

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- Hardware implementation of high-precision discrete Gaussian sampler.
 - Efficient implementation for small standard deviation
 - Storage is an issue for large standard deviation
- Implementation of LWE cryptosystem
 - Polynomial multiplier and discrete Gaussian sampler in pipeline
 - Sampler is slower than polynomial multiplier
 - Parallelization of sampler cores is possible

Thank You