# Analog Representations in Digital Arithmetic: A Review

Behrooz Parhami University of California, Santa Barbara

#### **About This Presentation**

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### **Analog Computing is Back!**

**Digital data has been replacing analog data for years:** From 1986 to 2007, share of digital stored data went from near-zero to 90% [1]; a decade later, digital is fully dominant

We now struggle with complexity and energy limitations: The solution seems to be approximate and analog computing

Hybrid digital/analog can give us the best of both worlds: Digital's higher latency and power used only when needed

Hence, multi-resolution representations are desirable: Combine fast, efficient low-precision computation with slower, energy-intensive high-precision computation



Analog Representation in Digital Arithmetic



### Hybrid Digital/Analog Representations

**Continuous-valued number system (CVNS):** A positional number system with analog digits

**Continuous-digit residue number system (CD-RNS):** Inspired by how positional info is represented in rat's brain

Here, we qualitatively compare the CVNS and CD-RNS representations

Race logic represents information as timing delays: Results derived based on relative signal propagation delays

**Space-time logic mimics spike-based neural computing:** Signal timing, as well signal magnitude, carries information

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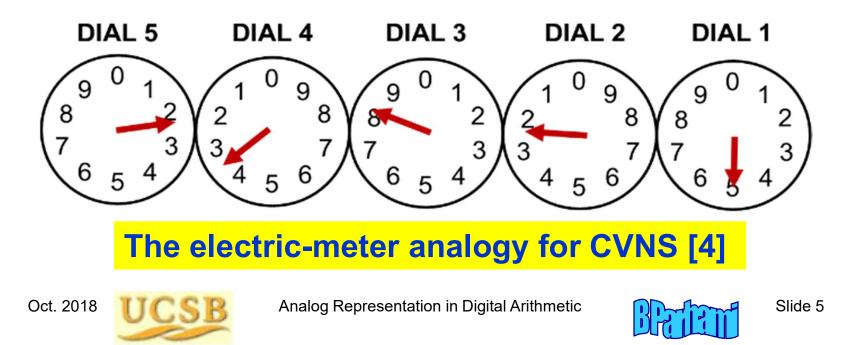


### **Continuous-Valued Number System**

#### **Positional representation with analog digits**

Contains a form of natural redundancy (digit values overlap) Leftmost analog digit ~2.38 tells us something about the next digits Overlaps make the representation/arithmetic more immune to noise

Can be based on unsigned or signed digits; e.g., (-5, 5)

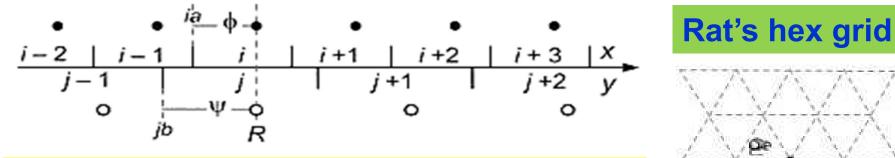


### **Continuous-Digit RNS Representation**

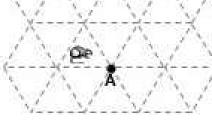
#### Modular arithmetic with continuous residues

Extension of RNS to non-integer moduli and residues Offers precision-range-robustness trade-offs More accurate residues widen the range and increase robustness

Of interest to neuroscientists: Rat's navigation system Rat uses "space cells" (absolute) and "grid cells" (relative) Can return to home position in the dark, without any visual cues



Localization with 2 grids in 1D space [6]



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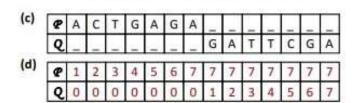


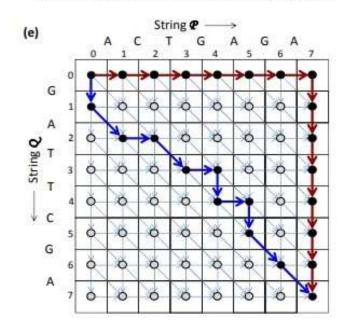
### **Time-Delay-Based Race Logic**

Not useful for general use (yet) Quite efficient in some domains

Example: String alignment [8] (as in DNA/protein matching) Closeness judged by "edit distance"

2D array of simple hardware cells  $O(m^2)$  hardware complexity

Paths represent alignments Horizontal, vertical, diagonal moves have different latencies Fastest path = Best alignment 



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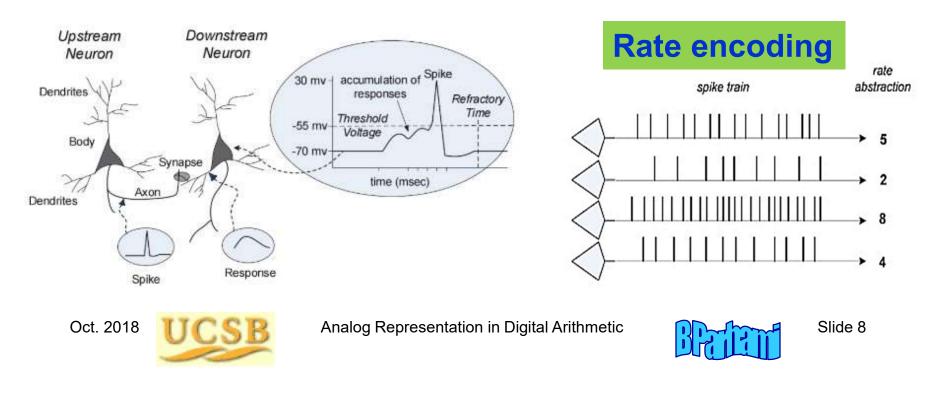


### **Brain-Like Space-Time Computing**

#### **Example: Transmitting one 8-bit byte representing** *n*

Binary: Delay = 9 slots; Average energy = 5 spikes Start/synch spike, followed by 8 spikes/no-spikes

Space-time: Average delay = 130 slots; Energy = 2 spikes Start/synch spike, followed by *n* no-spikes, then end spike

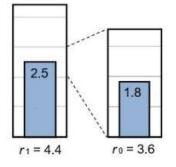


### **CVNS and CD-RNS Similarities**

**Two-level scheme:** Analog representation at the low (digit) level and digital interpretation at the high (inter-digit) level CVNS likely has performance edge in general applications

**Mixed-radix format:** CVNS is based on fixed integer radix, but extension to mixed and non-integer radices is possible

Example mixed-radix CVNS: Representation of 9.0 = 2.5 × 3.6 = 2 × 3.6 + 1.8



**Approximate computing:** Both CVNS and CD-RNS suited to low-precision, adaptive-precision, and lazy arithmetic





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### **CVNS and CD-RNS Differences**

**Word-level parallelism:** CD-RNS has greater affinity with parallel processing of digits in add/subtract/multiply

Input/output overheads: CVNS has simple/direct forward and reverse conversion processes (low-cost and low-energy) CD-RNS conversions are even more complex than RNS

**Noise immunity:** Consider 2-digit CVNS and CD-RNS CVNS range decreases quadratically with increased noise immunity CD-RNS range decreases linearly with increased noise immunity (cutting the radix *r* in half, versus using the equation  $\mu \epsilon \le \mu_0 \mu_1$ )

**Fault tolerance:** CVNS can be protected through coding CD-RNS has precision-robustness trade-off built in

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### **Conclusions and Future Work**

Analog computing is making a comeback and hybrid digital/analog computing is becoming more attractive

D/A computing can be combined at various levels:



Representation level, as in CVNS and CD-RNS Analog approximation, digital refinement Neuromorphic computing paradigm Multi-level combination methods

Future work and more detailed comparisons



Assessment of relative speeds in application contexts Quantifying cost and energy requirements Effects of radix and moduli selection Other D/A combination methods

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### **Questions or Comments?**

parhami@ece.ucsb.edu

http://www.ece.ucsb.edu/~parhami/

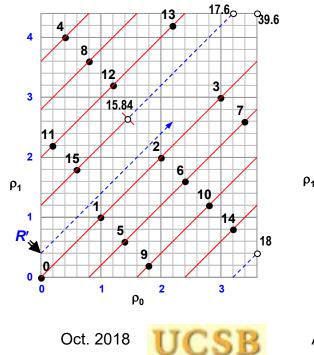
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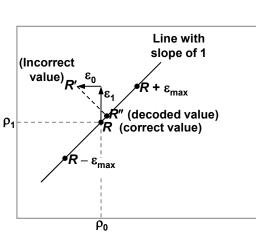
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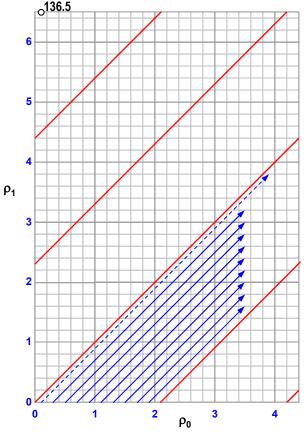
### **Range-Precision Trade-off in CD-RNS**

# Dynamic range is proportional to the product of moduli, divided by maximum error

Left diagram: Range extension beyond  $\prod m_i$ Middle diagram: Decoding error Right diagram: CD-RNS with  $m_1$  = 6.5,  $m_0$  = 4.4







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### **Forward/Reverse Conversion in CD-RNS**

#### **CRT-like methods of conventional RNS do not carry over**

Small errors in residues may be amplified through the conversion process A possible way out: View reverse conversion as nonlinear optimization Single-layer feedback network (see diagram on the bottom-right) Converges to the correct result within the *RC* time constant of the circuit

