

High-Dimensional Computing SDM Memory & Hyperscaling A session on the Cloud's missing AI component

Gil Russell WebFeet Research Inc.





"Science advances one funeral at a time."

Max Planck

"Randomness is the path of least presumption."

Pentti Kanerva





On the read to Arthfeld General Intelligence - What we have learned (so far)

Machine Learning ≠ Intelligence

Deep Learning Promise

• Has entered the trough of disillusionment

<u>Sequences</u> have become mandatory

• Requires hierarchical temporal capability

<u>Prediction</u> absolutely required

Fast learning required for anomalous conditions

Broad span solution

- Embedded Controllers through High-Performance Computing
- Cognitive Database compatibilities













Old Brain => Deep Learning (ANNs)

/// - A UDEMODITY IN UPANSIUM

Stimulus, Instinct & Response Low generalization Good at high probability inference Requires large amounts of training data Limited real time processing and incapable of real time predictive reasoning - brittle failure modes Autonomous Vehicle developers are rethinking their technology tracks





New Brain => *Hyperdimensional Computing*

Relates human intelligence to the properties of abstract mathematical spaces

Associates patterns and analogies – What if and Why?

Single pass training, High Noise Immunity High range of applicability, from toys to supercomputers

Can <u>predict</u> probable outcomes





Ayper-Dimensional Computing (HDC) Coogle: Visualizing High Dimensional Space

OU Brath' Memory

Artificial Neural Networks adoption of the Deep Learning algorithm (DL) has limited the development of algorithms that can deal with generality and the mapping of associations

DL designs rapidly adopted Parallel Arrays of Multiply-Accumulate (MAC) solutions using High Bandwidth Memory (HBM)

Memory is internal to the algorithm used and has no external object "association" tether

Dependence on a purely statistical algorithm with little reference to a memory framework is DLs weakest link bFeet cearch

New Brain" Memory Fabric

HDC effect is subtle

- Requires overlay of an Cognitive Object Data Base
- Node (Object) and Vector Edge represent either Hamming Distance SDM or Cosine Overlap in Hierarchical Temporal SDR to nearest and farthest relations
- Vector relationships are being incorporated into Cognitive Object Databases and Probabilistic Programming Languages





"Spared, Distributed Memory (SDM)"



Pentti Kanerva

Redwood Center for Theoretical Neuroscience "Sparse, Distributed Memory", MIT Press, 1988



"Hyperdimensional Computing: An Introduction to Computing in Distributed Representation with High-Dimensional Random Vectors", Pentti Kanerva, January 2009

https://redwood.berkeley.edu/









Jeff Hawkins

Founder: Redwood Center for Theoretical Neuroscience (2002), CEO Numenta (2005)



"On Intelligence, How a New Understanding of the Brain Will Lead to the Creation of Truly Intelligent Machines, 2004"

https://numenta.com/





New Brain - Key People Redwood Neuroseieneeinstitute (RND)

Pentti Kanerva Research Affiliate at the Redwood Neuroscience Institute

Dileep George Cofounder Vicarious AI Cofounder and CTO Numenta Bruno Olaufson Professor Helen Wills Neuroscience Institute Director and Cofounder, Redwood Center for Theoretical Neuroscience and Advisor to Vicarious AI

Jeff Hawkins Cofounder of Numenta, Redwood Center for Theoretical Neuroscience (2005) and Redwood Neuroscience Institute (2002)



http://www.rctn.org/wiki/History

DWOOD

12







https://www.wired.com/2014/04/vicarious-ai-imagination/

FILTIFE COUNTINE COMPUTING ALGINERAL - THEMAS

New computation models

WebFeet Research Research | Vision | Knowledge

Tailored architectures

Intrinsic device properties

Memory processing unit, can perform different tasks, data, storage, arithmetic, logic and neuromorphic computing using the same physical fabric that is programmable at the finest grain, the individual device level, without the need to move data outside the fabric.

Byper Inensional Computing Simple Block Diagram?

- HV Address space $(2^{10,000} = 1.9950631168807583848837421626 e+3010)$ • Associative Memory Processor Training + Experiential < 4TB
- Associative Memory Processor Training + Experiential < 4TB

Leverage unique structure & properties of VRRAM MAP kernels: demonstrated on 4-layer 3D VRRAMs In-memory HD computing enabled in 3D architecture

THE WEITTER WILLE CONTROL SU WITH A

In Ron-Volatile Memory - RURAW RAW

DDR4 Functionality

Low latency $CL \le 13.5 | 5 \text{ ns}$

Higher Density Roadmaps 8-32 | 4 Layers

Persistence 10 years | 300 years

Endurance 10¹² | 10¹⁵ PE

Zero Refresh (or very low overhead) | None

General and scalable model of computing with a well-defined set of arithmetic operations

- Fast and one-shot learning
- Auditable (explainability) opens entirely new segment opportunities
- Training sets reduced by orders of magnitude while still maintaining classification accuracy
- A *memory-centric architecture* with significantly parallelizable operations

Extremely robust against most failure mechanisms and noise

Predictive Analytics with high accuracy easily incorporated

Associative Base technology for the Cognitive Software Pyramid

ISSCC 2018 / SESSION 31 / COMPUTATION IN MEMORY FOR MACHINE LEARNING / 31.3

PULP-HD: Accelerating Brain-Inspired High-Dimensional Computing on a Parallel Ultra-Low Power Platform

4-core platform (1.5mm2, 2 mW)

Hyperdimensional Computing Associative Processor – NVDIMM-X Form Factor

288 Pin NVDIMM Form Factor

The Kanerva Partition (Database Intersect) In-Memory Hypothetical System Implementation

The Kanerva Partition (RT-PA Grouping) In-Memory Hyperdimensional Associative Processor (Hypothetical)

Next Up: 'Big Problems for Nervous Tissue'

WebFeet Research, Inc.

www.webfeetresearch.com

Alan Niebel, CEO

- 0.831.373.3303
- M.831.402.5754
- Niebel@webfeetresearch.com

Gil Russell, Principal Technology Analyst

- 0.510.589.9568
- F.510.489.1701
- Russell@webfeetresearch.com

