Lossless Medical Image Compression for Radiology





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Research Team

Research commenced in January, 2020 to build a data warehouse for training artificial intelligence models using distributed computing techniques.

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Dominick Romano - Founder, drainpipe.io - Heterogeneous Hierarchical Ontologies

- Rendering systems for video games
- Massively Parallel Artificial Intelligence



Felipe Carino, Jr. - Fmr. Chief Architect, Teradata
Researcher in Distributed Computing & Databases
US Patents - Awarded 8 U.S. Patents

- Validated work done at drainpipe as interim CTO

Introduction



Loading, storing, and visualizing large Neuro Informatics files (NII) commonly used in CT and MRI is costly and time consuming. To load the media, and store it for long term is extremely costly. To process the files, and transfer across systems is extremely time consuming. As more medical samples are accumulated and used to train AI Models, we must rethink how we store and process these files.

Representation Phases





Vectorizing Medical Imagery



[1] F. Cariño and W. Sterling, Parallel Strategies and Concepts for a Petabyte Multimedia Database Computer, *IEEE Parallel Database Techniques*, 1998.



Hilbert Symbolics

Using symbolics generated from hilbert space for lossless medical imagery compression in parallel computing strategies



Hilbert segmented images are recursively processed, this is distributed across multiple threads and systems for each hilbert cube.

Hilbert Array (Buffer)	Array Length 16
[0] [1] [2]	[3]
	Max Hold: 160 Items
	Max Siz (ex. 10
-ile 🚽	
Header	
FeatureCount : Int MaxSize : Int Vec3d Min	
MaxLength : Int Vec3d Max	
[0] Cube Count Item Count Value	
[1] Cube Count Item Count Value	
[2] Cube Count Item Count Value	
]
[15] Cube Count Item Count Value	
Bucket Allocate	d Size: 10
Index : Int Count : Int	
Vec 3d (X, Y, Z : Double) Value : Double	
	<u> </u>
Vec 3d (X, Y, Z : Double) Value : Double	
Vec 3d (X, Y, Z : Double) Value : Double	
Vec 3d (X, Y, Z : Double) Value : Double	
Bucket	
Bucket	
Bucket	



NIST Medical Databank



8.4 INTERCONNECTION NETWORK TOPOLOGIES

The following illustrations and discussion^{12,29,30,32} describe interconnection network topologies and performances. Figure 8 pictorially shows the topology for 2D-Mesh, Hypercube, Crossbar, BANYAN, and the Bynet.

Table 1 is a qualitative comparison of network topologies. Fable 2 provides a quantitative practical comparison using 64 nodes as an example. SDC-OMEGA,¹⁶ EDS-DELTA,^{20,34} MESHNET,³⁵ and iPSC/860-PARAGON³⁶ are other interconnection networks that have been designed since the survey was written.²⁹ Our description above shows the thought processes and rationale behind the Bynet design choices.



* Bynet based on 8x8 switch nodes; picture uses 4x4 nodes.

Figure 8. Interconnection network topologies.

Figure 5. Bynet network topology.

[2] Industrial Database Supercomputer Exegesis: The DBC/1012, The NCR 3700, The Ynet, and The Bynet, Chapter 8

Performance Benchmarks



References

[1] F. Cariño and W. Sterling, Parallel Strategies and Concepts for a Petabyte Multimedia Database Computer, *IEEE Parallel Database Techniques*, 1998.

[2] Felipe Cariño Jr., Warren Sterling, and Pekka Kostamaa, Industrial Database Supercomputer Exegesis: The DBC/1012, The NCR 3700, The Ynet, and The Bynet, Chapter 8, pp 9 11, PARLE Lecture Notes, 1992



Thank you

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