Reducing Memory Access Latencies using Data Compression in Sparse, Iterative Linear Solvers All-College Thesis Defense

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Motivation

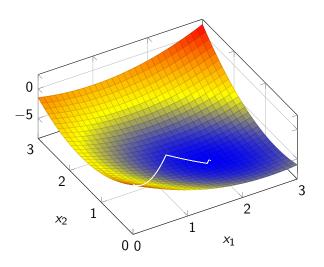
- Solving sparse linear systems used in many computations
- Iterative solvers are used
- Spend most of the time fetching data

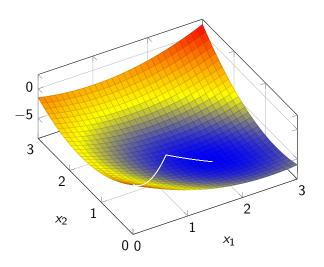
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- If **A** is symmetric, then $\nabla f(\vec{x}) = \mathbf{A}\vec{x} \vec{b}$
- $-\nabla f(\vec{x})$ is in direction of maximal decrease





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 - Stores the column index and value for each nonzero entry
- 3 compressible data structures
 - Vector Values
 - Matrix Indices
 - Matrix Values

Compression Methods

- Mixed Floating Point Precision
- SZ Compression
- Elias Gamma and Delta Codings
- ZFP Compression
- Huffman Coding
- Op Code Compression

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Mixed Floating Point Precision

- Single Precision takes half the storage space
- Only 6-9 digits instead of 15-17 digits
- Certain vectors can be lower precision without slowing convergence
 - Retains result accuracy

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- Available prediction functions are based on the data
- Compression rate is highly dependent on local patterns in the data

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 - Very effective for small integers
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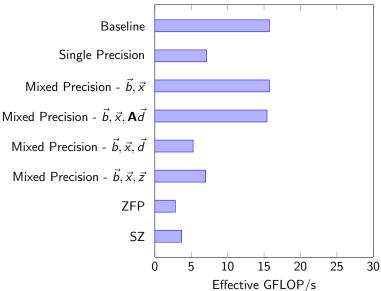
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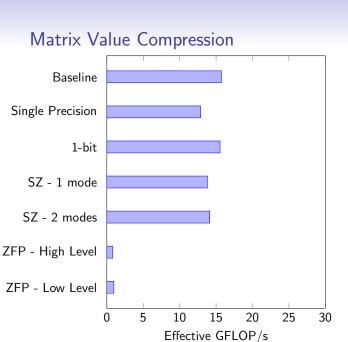
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 - Storing the difference from the previous index reduces the size of values
- Elias Delta Coding is similar, but uses Gamma coding for the length
- Compression rate is only dependent on the magnitude of the values

Timing Results

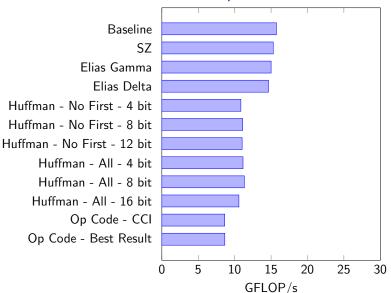
- 60 processes with 96³ rows each
 - 53,084,160 total rows
- A 20-core, 2.2GHz, Intel Broadwell head node
- Plus five 8-core, 1.7GHz Intel Broadwell nodes
- MPI communication



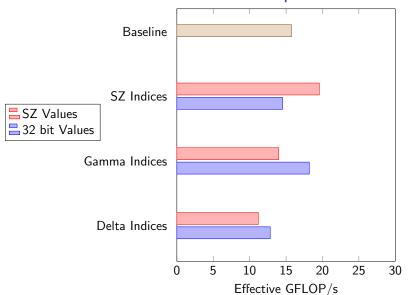




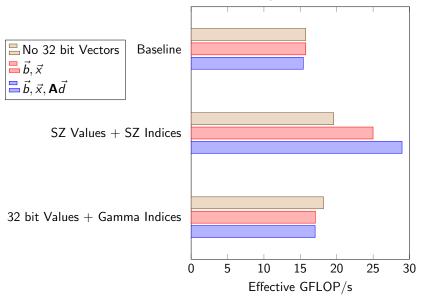
Matrix Index Compression



Matrix Value and Index Compression



Vector and Matrix Compression



Conclusion

- Iterative, sparse linear solvers are memory access bound
- Compressing key data structures provided an 84% increase in performance