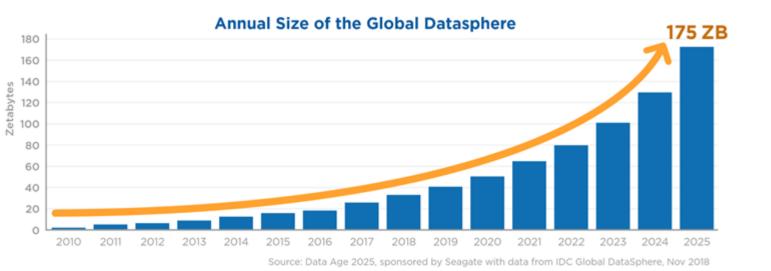


GSST High Throughput Parallel String Decompression on GPU Robin Vonk





Data Growth

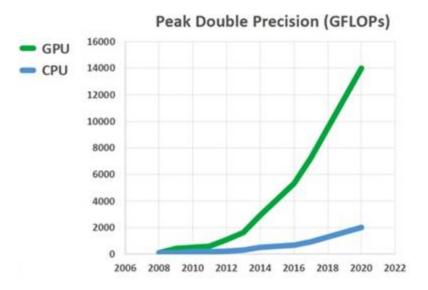




The volume of data doubles every 2 years

The need for faster processing keeps growing

Graphics Processing Unit



Source: https://www.nextplatform.com/2019/07/10/a-decade-of-accelerated-computing-augurs-well-for-gpus/

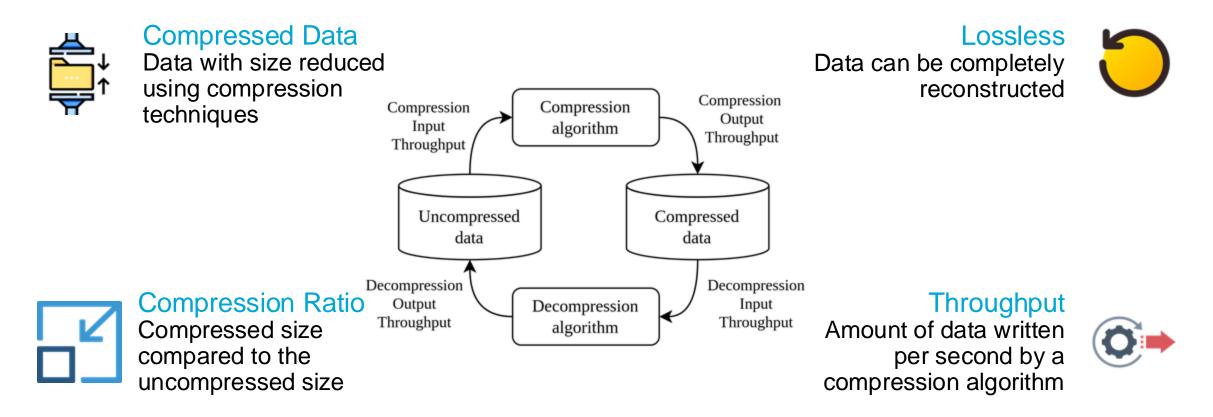


Source: https://lenovopress.lenovo.com/lp1734-thinksystem-nvidiaa100-pcie-40-gpu





Lossless Compression





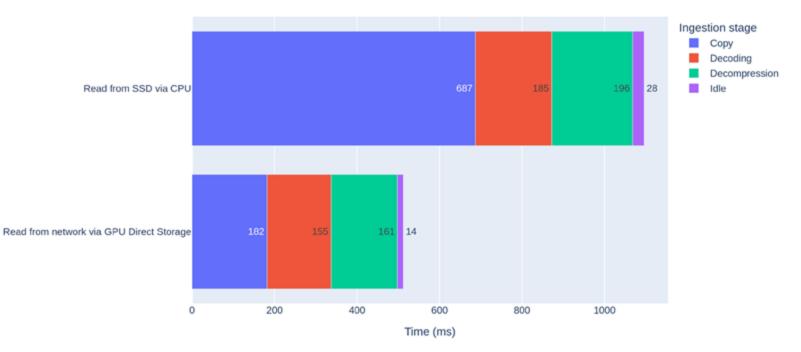
Data Ingestion

The process of preparing stored data for processing

As data transfers are getting faster, decompression time becomes more significant

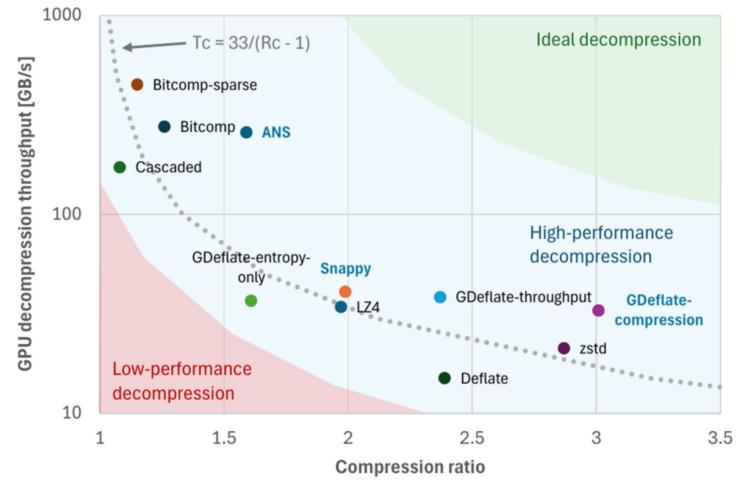
On fast network storages, decompression can take as long as the transfer itself

Storing data uncompressed can be faster than using slow decompression





GPU Lossless String Decompression



- Data ingest throughput depends on
 - Compression ratio
 - Decompression throughput

TUDelft

How to increase data ingestion throughput of string data on GPUs?



Topics



FSST Compression 02

GSST Decompression Format 03

Memory Access Optimizations 04

Results and Conclusions





FSST Compression



FSST Compression

Fast Static Symbol Table

Unmatched single threaded decompression throughput

Compression

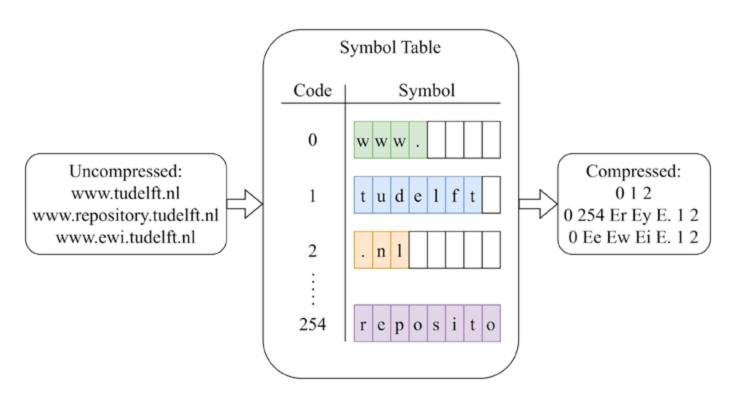
- 1. Scan input for repeated **symbols**
- 2. Replace symbols with codes

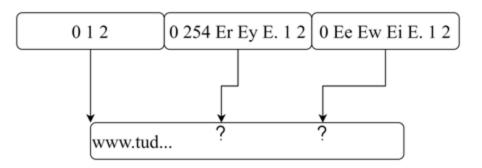
Decompression

1. Replace codes with symbols

Only one sequential problem:

Where to write the output data?





GPU Hardware



Parallelism 6912 Cores over 108 streaming multiprocessors



High Bandwidth Memory 2 TB/s memory throughput



Source: https://images.nvidia.com/aem-dam/en-zz/Solutions/data-center/nvidia-ampere-architecture-whitepaper.pdf

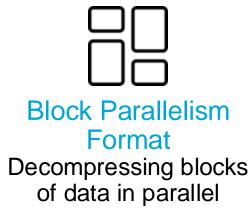


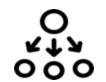


GSST Decompression Format

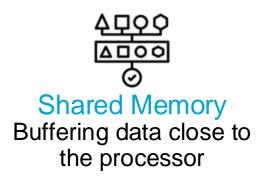


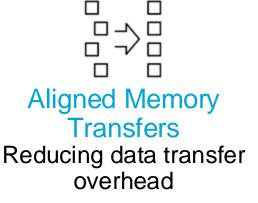
GSST Decompression





Split Parallelism Format Decompressing a single block in parallel







Block Parallelism

Divide the input data over independent compressed blocks (tiles)

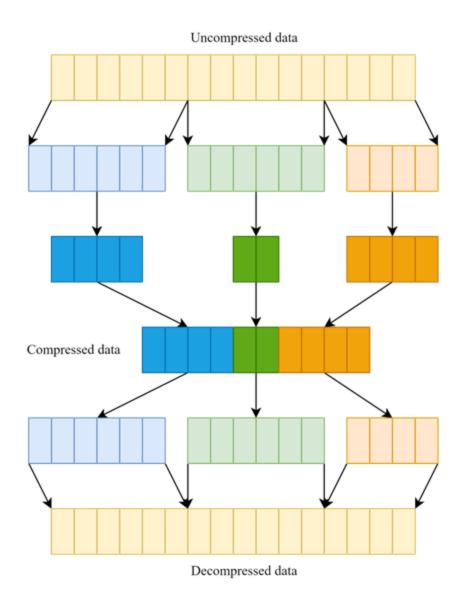
Stores the compressed and uncompressed size of each block

Pros

• Parallelism equal to the number of blocks

Cons

- Metadata lowers compression ratio
- Doesn't utilize full processing power of the GPU



TUDelft

Splits Parallelism

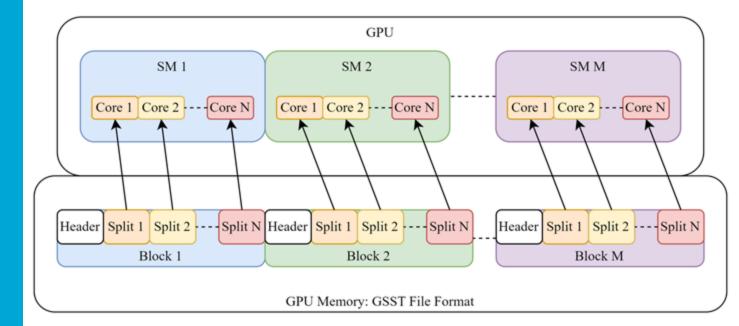
Divide a block into splits to allow parallelism within a block

Pros

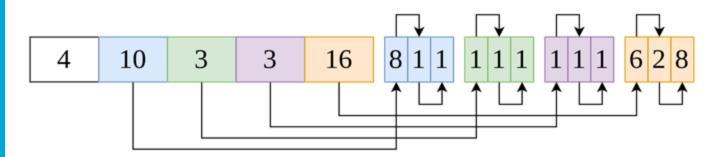
Parallelism equal to the number of splits

Cons

Metadata lowers compression ratio



number of splits	split	split	split	split	8	1	1	1	1	1	1	1	1	6	2	8
splits	location	location	location	location	0	1	1	1	1	1	1	1	1	V	2	9

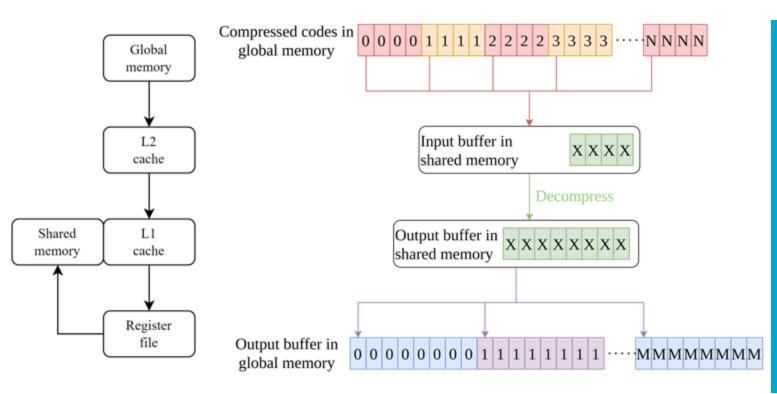






Memory Access Optimizations





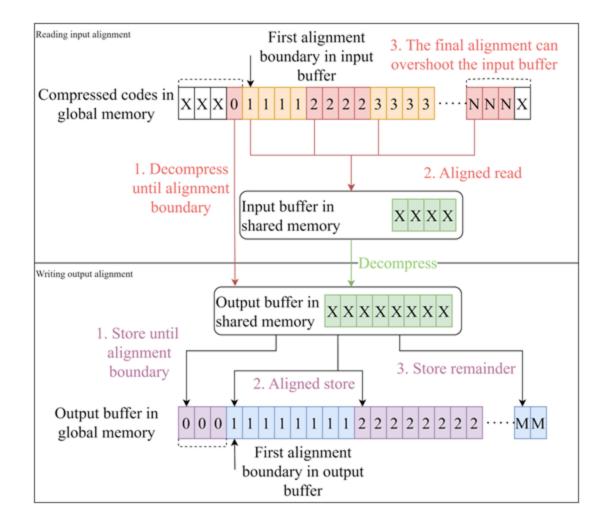
Shared Memory

Shared memory allows manually storing data in L1 cache Shared allows efficiently sharing data between threads

- Store the symbol table in shared memory
- Buffer input and output data in shared memory



Memory Alignment



Using aligned transfers

- 1. Find the first aligned location in the buffer
- 2. Use slow data transfers until the aligned location
- 3. From the aligned location, use fast aligned transfers
- 4. Handle the remainders

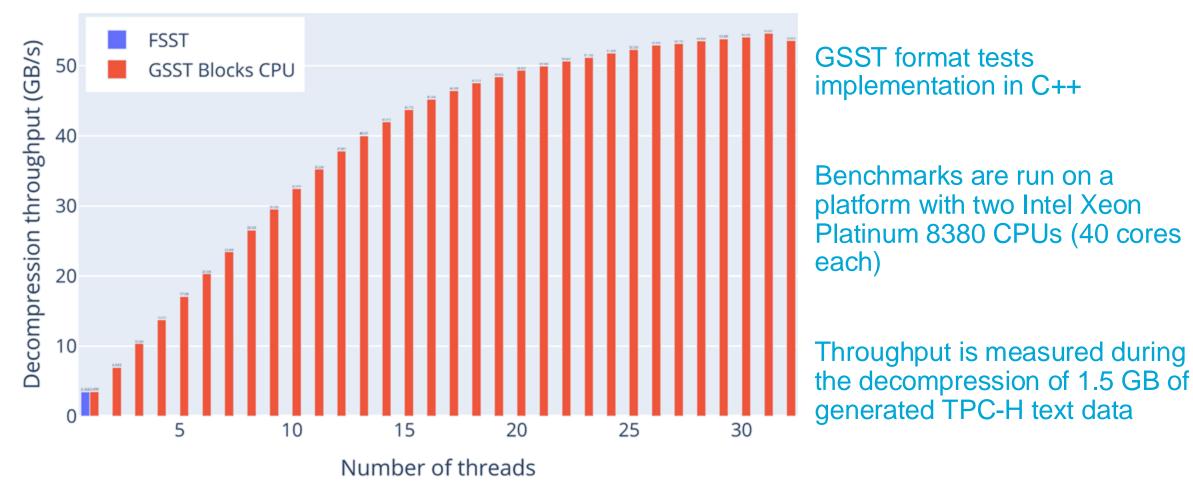




Results and Conclusions

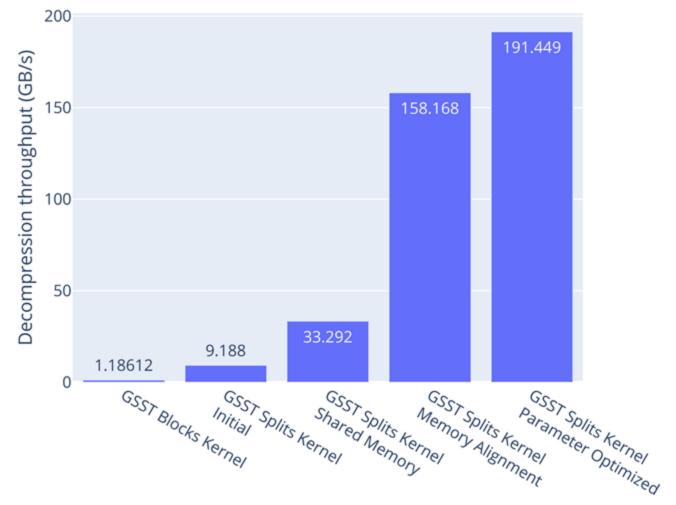


GSST CPU Performance



TUDelft

GSST Implementations Comparison



All optimizations are implemented as CUDA kernels

Benchmarks are run on NVIDIA A100 80GB GPU

Throughput is measured during the decompression of 10 GB of generated TPC-H text data

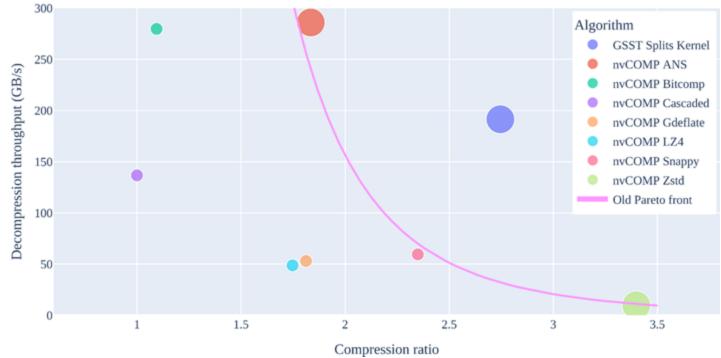


GSST Performance

Decompression throughput of **191 GB/s** with compression ratio of **2.74**

ANS has a 49% higher throughput, but GSST 49% higher compression ratio

Zstd has a 23% higher compression ratio, but GSST has 18 times higher throughput



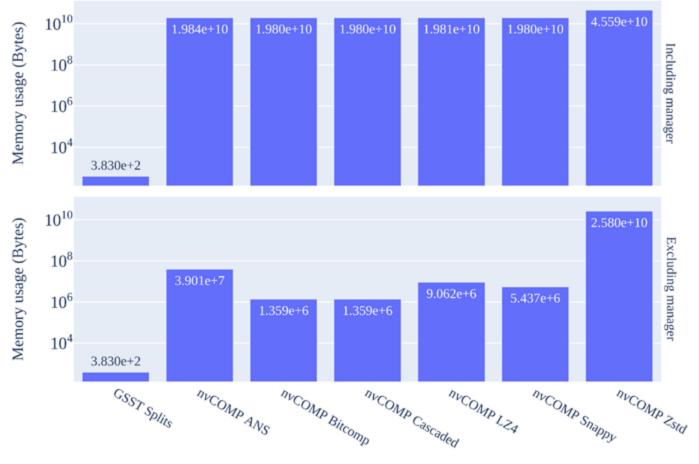


Memory Usage

State-of-the-art GPU compression in nvCOMP is extremely memory inefficient

3,500 times less memory usage compared to nvCOMP's most efficient decompression Bitcomp

67 million times less memory usage compared to nvCOMP's most memory-intensive decompression Zstd



Decompression algorithm





Thank You!

GSST: High Throughput Parallel String Decompression on GPU | Robin Vonk

31-3-2025