

Lossless Data Compression on GPUs

Ritesh Patel

University of California, Davis

Jason Mak
University of California, Davis

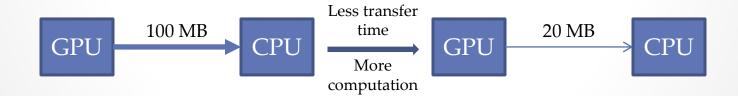


Motivation

Data storage

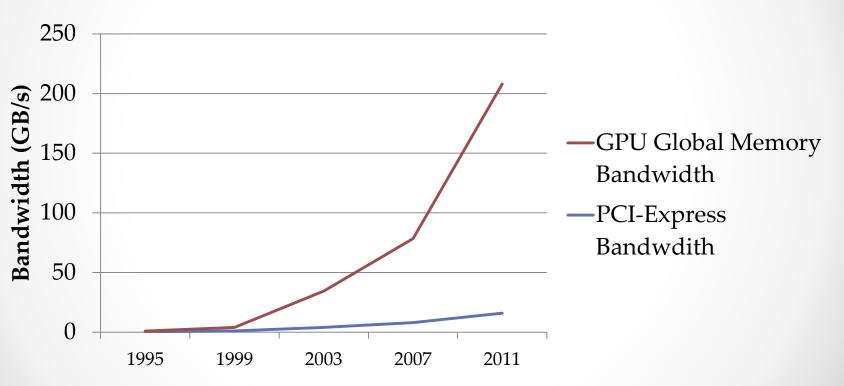


- Computation vs. Data Transfer
 - o Is compress-then-send worthwhile?





PCI-E vs. GPU bandwidth





Topics Covered

- Related Work
- Three Algorithms in the Compression Pipeline
 - 1. Burrows-Wheeler Transform
 - 2. Move-to-Front Transform
 - 3. Huffman Coding
- Results
- Future Work

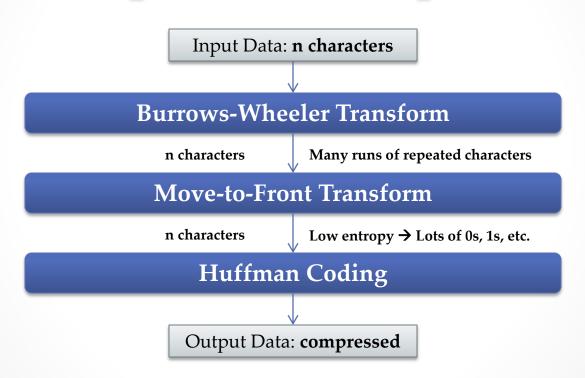


Related Work

- Domain-specific compression
 - Floating-point data compression
 - Texture compression
- Parallel bzip2 (pbzip2)
 - Uses pthread library
 - bzip2 not data-parallel-friendly

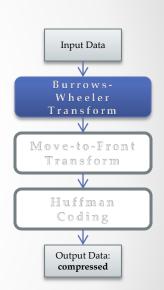


Compression Pipeline





- Transforms a string and gives has many runs of repeated characters
- Same characters get grouped





- Transforms a string and gives has many runs of repeated characters
- Same characters get grouped



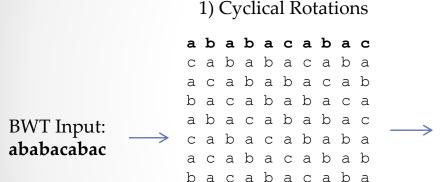




1) Cyclical Rotations







abacabacab

babacabaca



2) Sort





Sorted Cyclical Rotations

```
a b a b a c a b a c
a b a c a b a b a c
a b a c a b a c a b
a c a b a c a b
a c a b a c a b
a c a b a c a b a c
b a c a b a c a b a c
b a c a b a c a b a c
b a c a b a c a b a
c a b a c a b a c a
b a c a b a c a b a
c a b a c a b a b
a c a b a c a b a
```

Index: 0

BWT: ccbbbaaaaa





BWT with Merge Sort

- Sorting n strings → each n characters long
 - o **n** = 1 million per block in our implementation
- Radix sort not a good fit
- Break ties on the fly





BWT with Merge Sort

- Sorting n strings → each n characters long
 - o **n** = 1 million per block in our implementation
- Radix sort not a good fit
- Break ties on the fly
- String sorting on GPU
 - o Non-uniform

Pair 1: 7 ties

Pair 2: 0 ties

 a
 b
 a
 c
 a
 b
 a
 c
 a
 b

 a
 c
 a
 b
 a
 c
 a
 b
 a
 c
 a
 b

 a
 c
 a
 b
 a
 c
 a
 b
 a
 c
 a
 b

 b
 a
 c
 a
 b
 a
 c
 a
 b

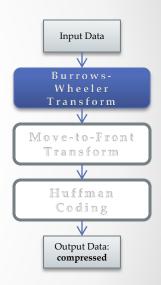
ababacabac

abacaba bac

Pair 3: 4 ties

c a b a b a c a b a **c a b a**

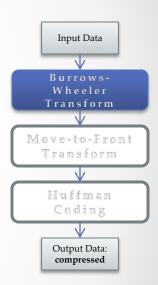
bacabacaba





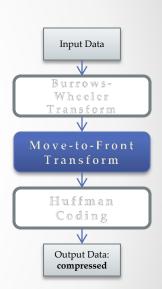
BWT with Merge Sort

- Sorting n strings → each n characters long
 - o **n** = 1 million per block in our implementation
- Radix sort not a good fit
- Break ties on the fly
- String sorting on GPU
 - o Non-uniform
 - o Parallelize BWT with string sorting algorithm based on merge sort
 - Currently fastest string sort on GPU





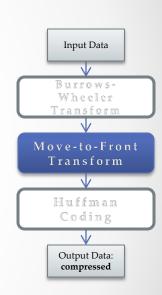
- Exploit clumpy characters from BWT
 - o BWT: ababacabac → ccbbbaaaaa
- Improves effectiveness of entropy encoding





- Each symbol in the data is replaced by its index in the list
 - o Initial list: ...abcd... (ASCII)
- Recently seen characters are kept at front of the list
 - Long sequences of identical symbols replaced by many zeros

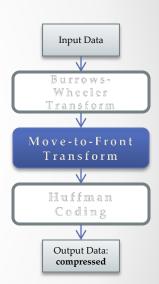






InputMTF ListOutputccbbbaaaaa...abc... (ASCII)[99]

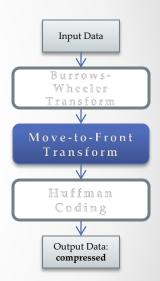
• 'c' occurs at 99th index → Output '99'





InputMTF ListOutputccbbbaaaaac...ab... (ASCII)[99]

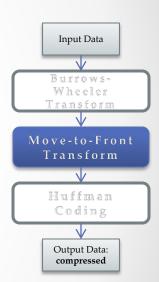
- 'c' occurs at 99th index → Output '99'
- Move 'c' to front of the MTF list
- Shift all previous elements to the right





Input	MTF List	Output
ccbbbaaaaa	c ab	[99 ,0]

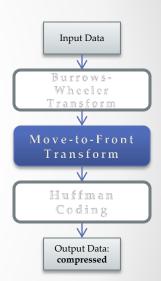
• 'c' occurs at 0^{th} index \rightarrow Output '0'





Input	MTF List	Output
ccbbbaaaaa	c ab	[99 ,0]

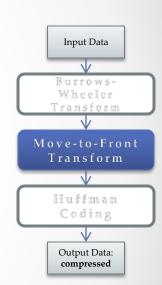
- 'c' occurs at 0^{th} index \rightarrow Output '0'
- 'c' already at front, no shifts





Input	MTF List	Output
ccbbbaaaaa	c ab	[99 ,0]

- 'c' occurs at 0^{th} index \rightarrow Output '0'
- 'c' already at front, no shifts
- Repeat for all elements





Iteration

ccbbbaaaaa

MTF List

...abc... (ASCII)

c...ab...

c...ab...

bc...a...

bc...a...

bc...a...

abc...

abc...

abc...

abc...

Transformed String

[99]

[99**,0**]

[99**,**0**,99**]

[99,0,99,**0**]

[99,0,99,0,<mark>0</mark>]

[99,0,99,0,0,**99**]

[99,0,99,0,0,99,**0**]

[99,0,99,0,0,99,0,**0**]

[99,0,99,0,0,99,0,0,<mark>0</mark>]

[99,0,99,0,0,99,0,0,0,0]





Parallel MTF

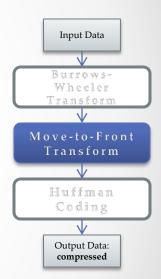
- MTF appears to be highly serial
 - Character-by-character dependency





Parallel MTF

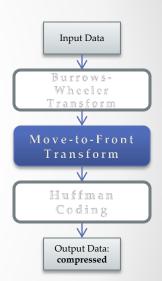
- MTF appears to be highly serial
 - Character-by-character dependency
- Goal: generate multiple MTF lists at different indices
 - Compute MTF output in parallel





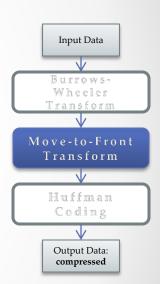
Parallel MTF

- MTF appears to be highly serial
 - Character-by-character dependency
- Goal: generate multiple MTF lists at different indices
 - Compute MTF output in parallel
- How to parallelize?
 - Two key insights
 - 1. Generate partial MTF list of a substring s
 - 2. Combine two adjacent partial MTF lists (using "concatenation")





BADADDDACCCBBAAA





BADADDDACCCBBAAA

End of part 1

MTF List

A D B





Part 1 Part 2

BADADDDACCCBBAAA

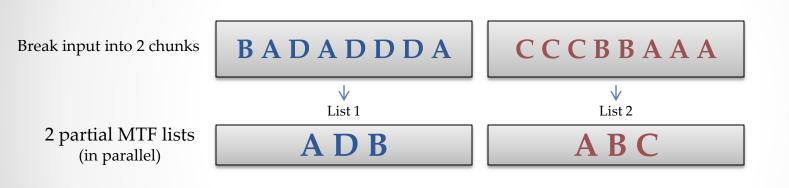
Final MTF List

End of part 2

ABCD

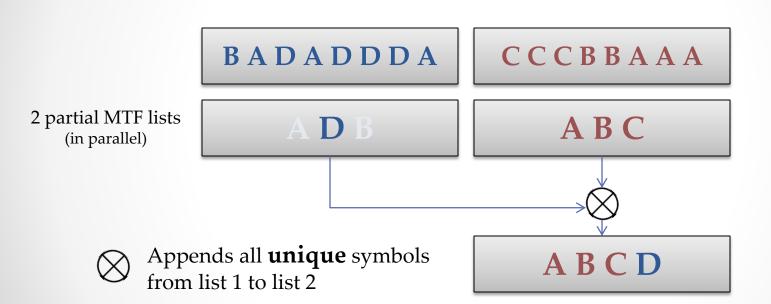


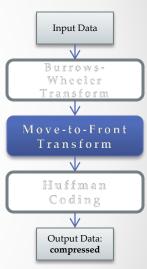




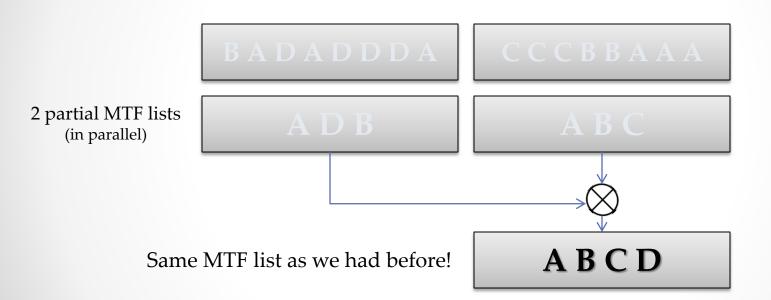


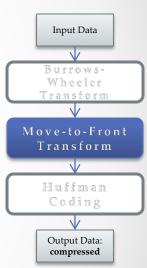




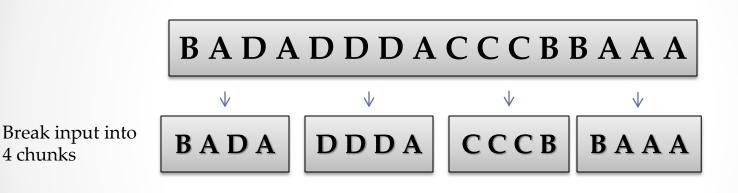






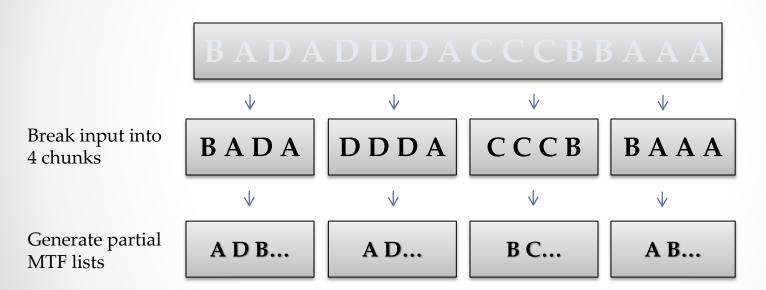


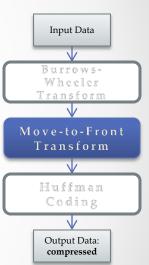




4 chunks









A D B

A D

B C

A B

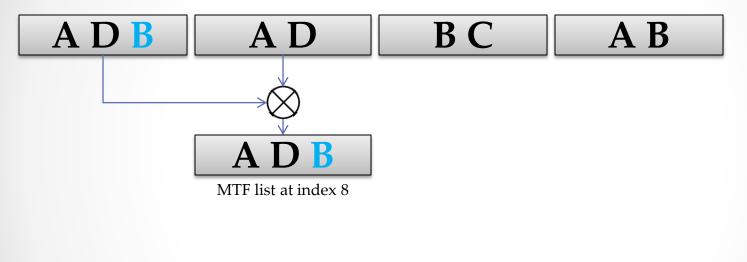
MTF list at index 4

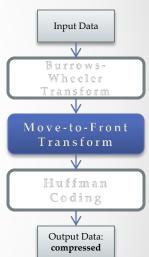




Operator → Appends all unique symbols to current list



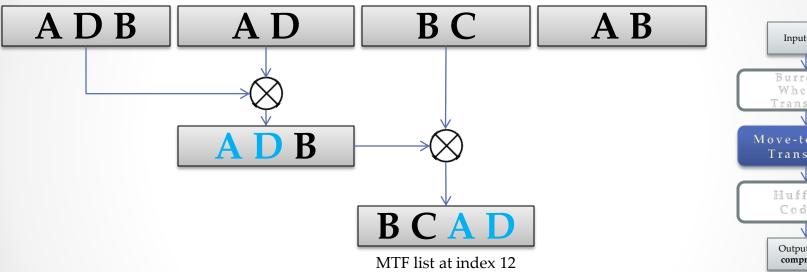






Operator → Appends all unique symbols to current list



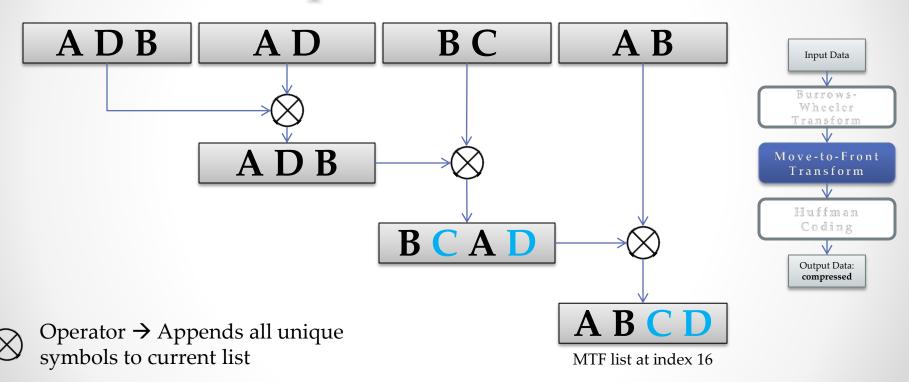




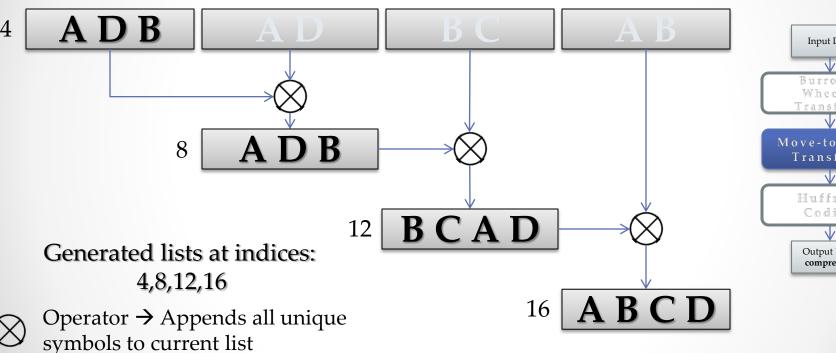


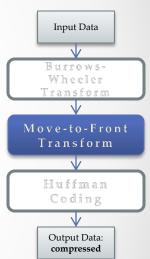
Operator → Appends all unique symbols to current list



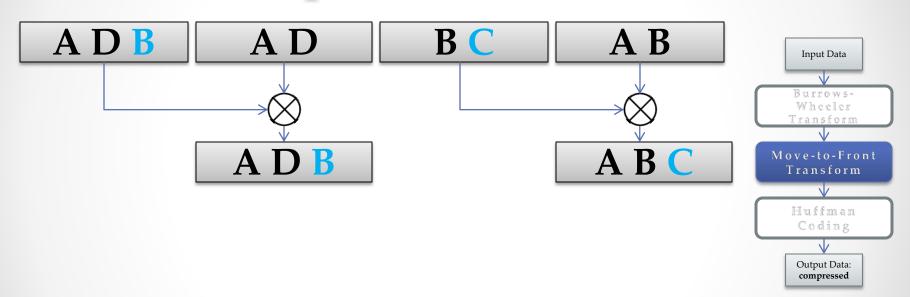


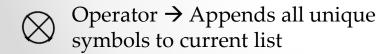




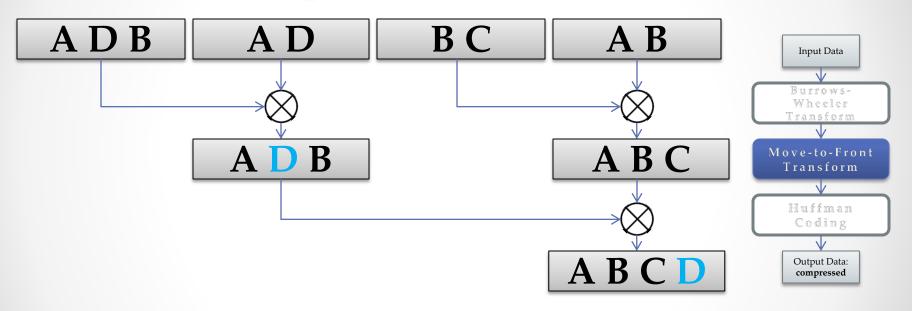


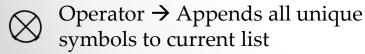




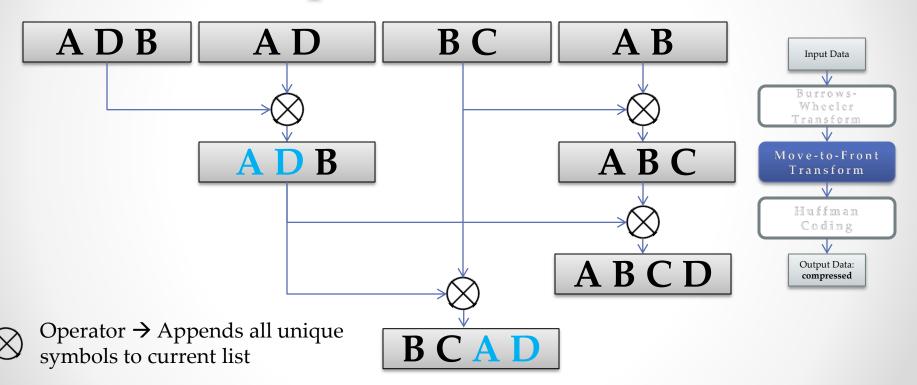




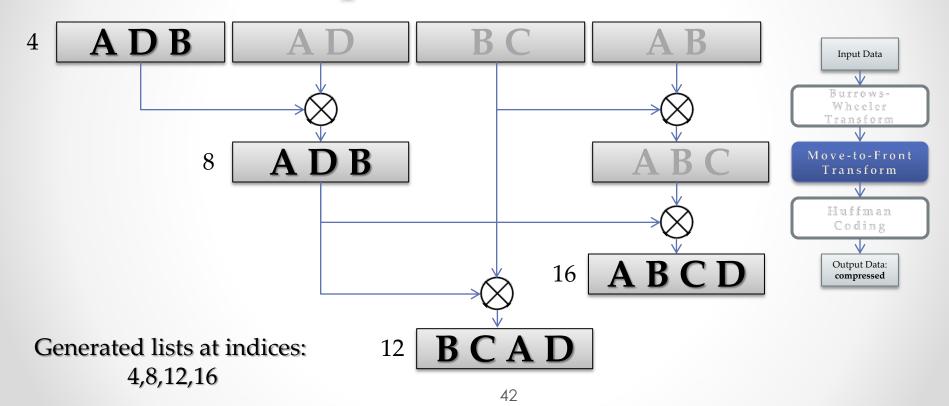




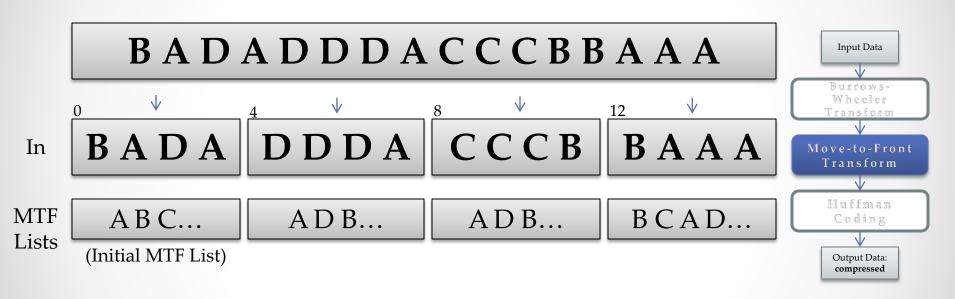




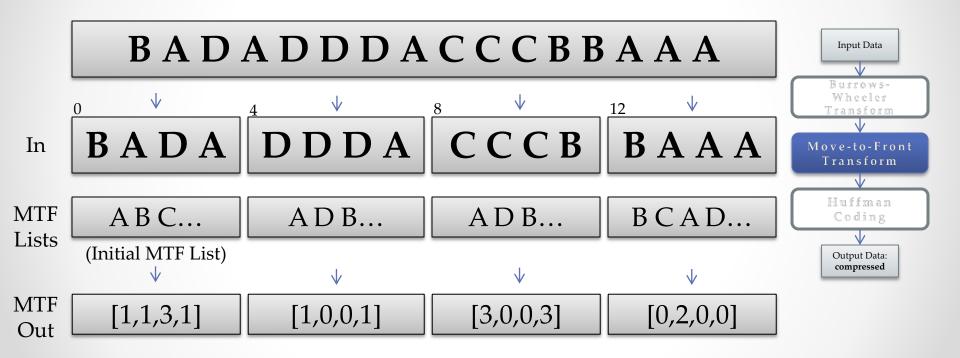






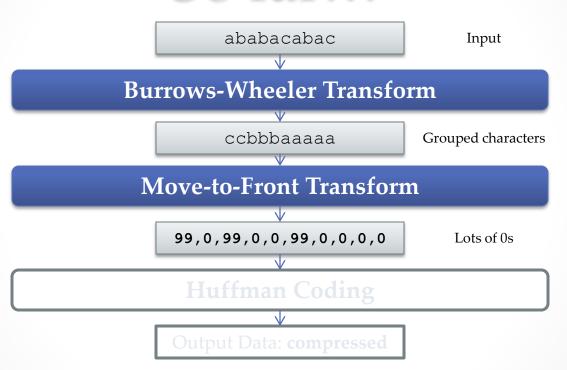






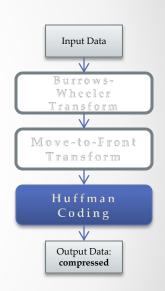


So far...



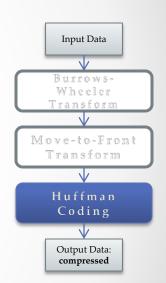


- Final stage that performs actual compression
- Replace each character with a bit code
- Characters that occur more often get shorter codes
- Huffman Coding is more effective with repetitive data (after MTF)



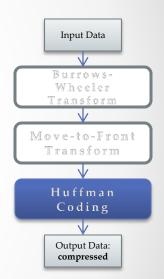


"HUFF" → 00 01 1 1 32 bits 6 bits





- Huffman Stages:
 - 1. Generate 256-bin histogram
 - 2. Build Huffman tree
 - 3. Replace characters with codes

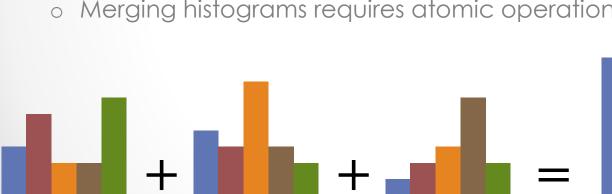


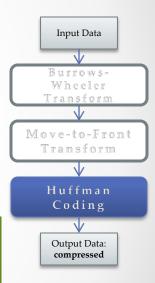


Huffman Histogram

- Histogram
 - Build a 256-entry histogram to count characters
 - o To do this on the GPU, divide the input data among threads with each thread maintaining its own histogram

Merging histograms requires atomic operations



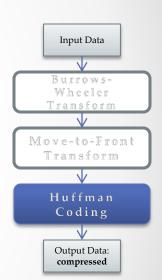




Huffman Tree

Huffman Tree Algorithm

- Remove two lowest counts, combine to form composite node
- Composite node "inserted" into histogram with combined counts
- o Each step is dependent on the previous step
- Parallelize finding the two lowest counts (maximum 256-way parallelism)





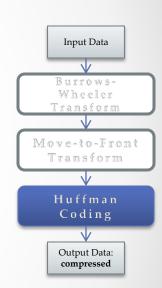
Histogram

H: 1

U: 1

F: 2

"HUFF"



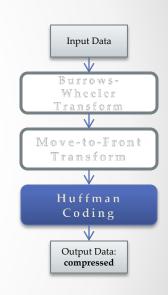


Histogram

H: 1

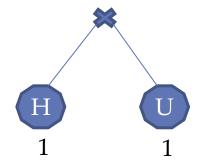
U: 1

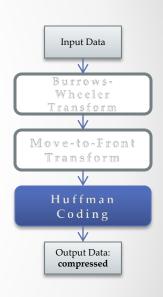






<u>Histogram</u>

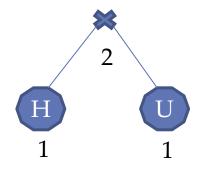


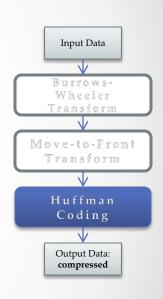




Histogram

H+U: 2

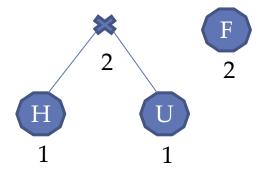


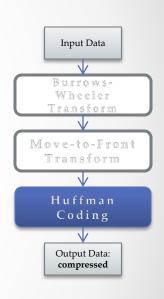




Histogram

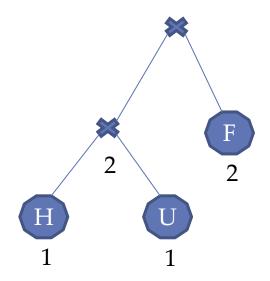
H+U: 2

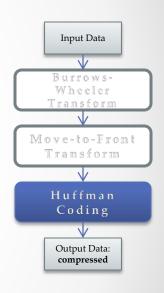






Histogram (empty)





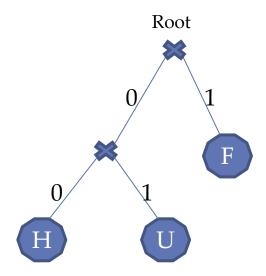


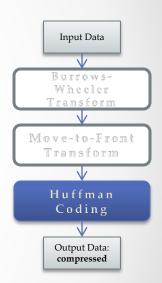
Left $\rightarrow 0$ Right $\rightarrow 1$

Huffman Codes

H: 00

U: 01





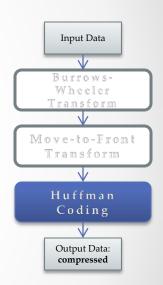


- Huffman coding assigns prefix codes
- No codes share the same prefix

Huffman Codes

H: 00

U: 01





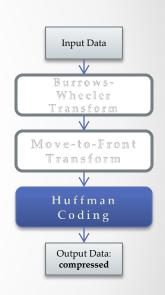
Huffman Codes

H: 00

U: 01

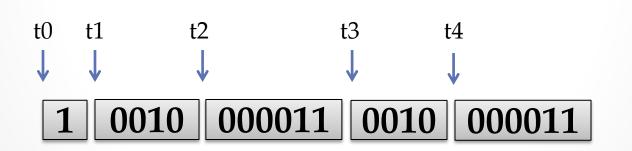
F: 1

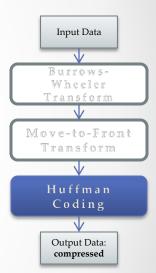
"HUFF" → 000111
32 bits 6 bits





- Replace characters with codes
 - To do in parallel, divide input among threads
 - Hard to do on the GPU because codes are variable-length
 - Each thread must calculate the correct offset to begin writing its code







- GPU vs. Bzip2
 - o Overall
 - GPU 2.78x slower
 - o BWT
 - GPU 2.9x slower (91% of runtime)
 - o MTF + Huffman
 - GPU 1.34x slower



Benchmark results

File	Compress Rate	BWT Sort Rate	MTF+Huffman Rate	Compress Ratio Compressed Size Uncompressed Size
(Size)	(MB/s)	(Mstrings/s)	(MB/s)	
Text	GPU: 7.37	GPU: 9.84	GPU: 29.4	GPU: 0.33
(97 MB)	bzip2: 10.26	bzip2: 14.2	bzip2: 33.1	bzip2: 0.29
Source Code	GPU: 4.25	GPU: 4.71	GPU: 44.3	GPU: 0.24
(203 MB)	bzip2: 9.8	bzip2: 12.2	bzip2: 48.8	bzip2: 0.18
XML	GPU: 1.42	GPU: 1.49	GPU: 32.6	GPU: 0.19
(151 MB)	bzip2: 5.3	bzip2: 5.4	bzip2: 69.2	bzip2: 0.10

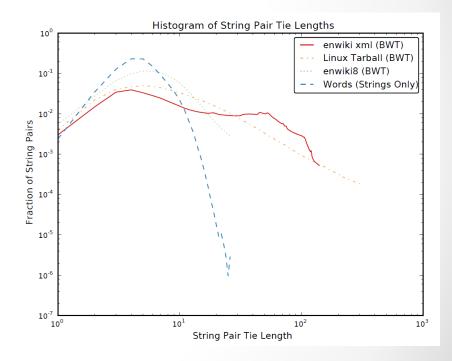
Worst performance during string sort and overall

Best amount of compression



BWT Performance

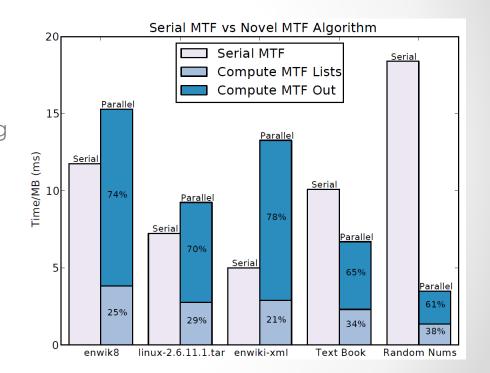
- o GPU 91% of runtime
- o Bzip2 81% of runtime
- o Tie Analysis:
 - Amount of compression is data dependent
 - Better compression leads to longer ties and poor performance





MTF Performance

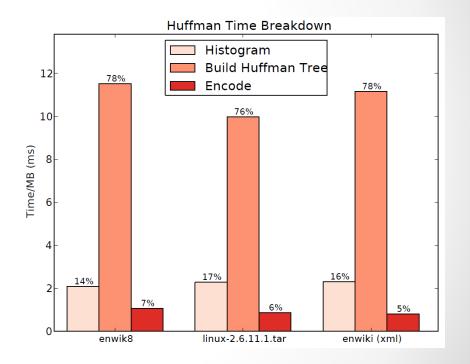
- Majority of runtime during character replacement
- Moving element to front, shifting all other elements
- Thread divergence





Huffman Performance

- Majority of runtime during Huffman tree building
- 256-way parallelism is inadequate for the GPU





Decompression

Reverse Burrows-Wheeler transform

- Much faster than forward BWT
- Requires only 1 character-sort
- Similar to linked-list traversal
- Preliminary implementation is an extension of Hillis and Steele's parallel linked-list traversal algorithm



Decompression

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Reverse Move-to-Front transform

- Parallel approach uses similar algorithm as forward-MTF
 - Generate partial MTF lists
 - Combine adjacent lists



Decompression

- Reverse Burrows-Wheeler transform
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 - Preliminary implementation is an extension of Hillis and Steele's parallel linked-list traversal algorithm
- Reverse Move-to-Front transform
 - Parallel approach uses similar algorithm as forward-MTF
 - Generate partial MTF lists
 - Combine adjacent lists
- Huffman decoding
 - Decode each encoded block in parallel



Future Work

- Explore other GPU-based string sorts that can better handle long strings with many ties
- Develop a Huffman tree building algorithm that has more than 256-way parallelism
- Overlap GPU compression and PCI-Express data transfer



Conclusion

- Implemented parallel lossless data compression on the GPU
- Parallelized BWT, MTF, and Huffman Coding
 - Developed a novel algorithm for MTF
 - BWT string sort contributes to the majority of runtime
- Implementation is slow but may be better suited for future GPU architectures and other parallel environments



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