# Reordering sparse matrices for fast multiplication on tensor architectures

**Github repository:** 

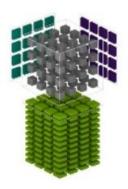
https://github.com/LACSFUB/SPARTA

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### **Motivation**

**Tensor cores** are good at multiplying dense arrays

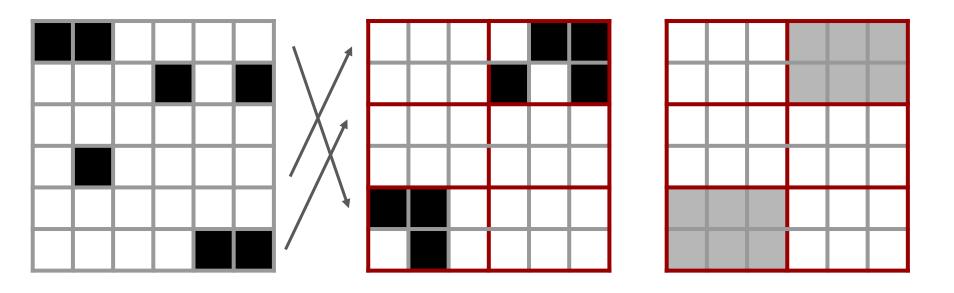
But what about **sparse matrices** multiplication?



Should we use sparse-specific data structures (CSR) and algorithms (cusparse)? We can't benefit of the tensor architecture

Should we treat the sparse matrix as a dense one and use tensor architectures? We end up multiplying a lot of zeros

### Our solution



ORIGINAL REORDERED BLOCKED

### **Applications**

Reordering takes time

If the same matrix is multiplied repeatedly, we can absorb the cost

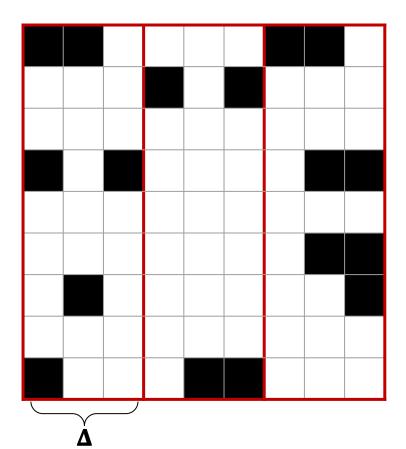
- graphs
- neural networks
- matrix collections

### The reordering algorithm

Find similar rows and group them together in blocks

- 1) Partitioning the columns
- 2) Grouping by hash
- 3) Grouping by similarity

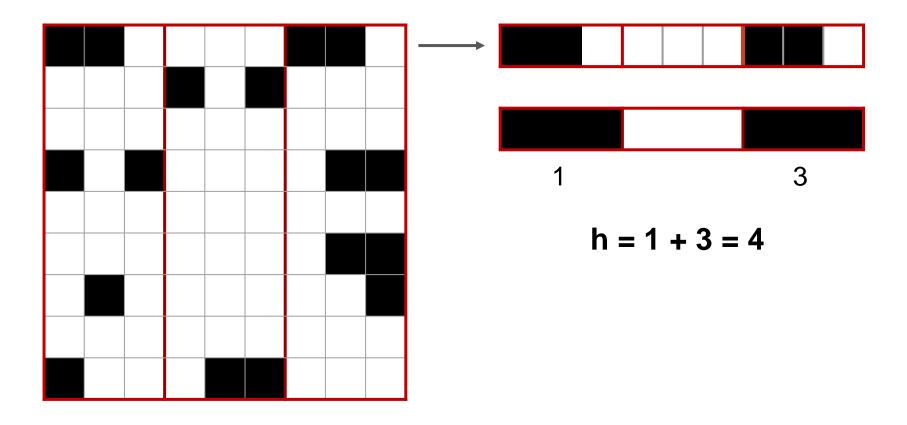
### The reordering algorithm – partitioning the columns



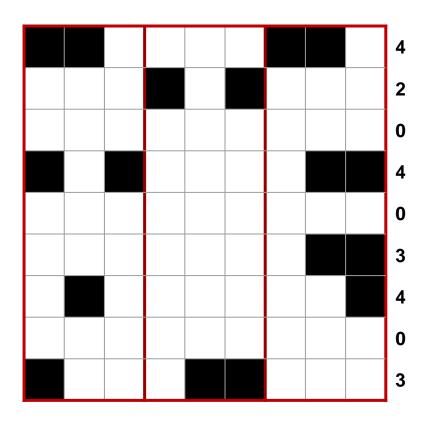
Choice of  $\Delta$  may depend on

- architecture
- previous knowledge of the matrix
- ...

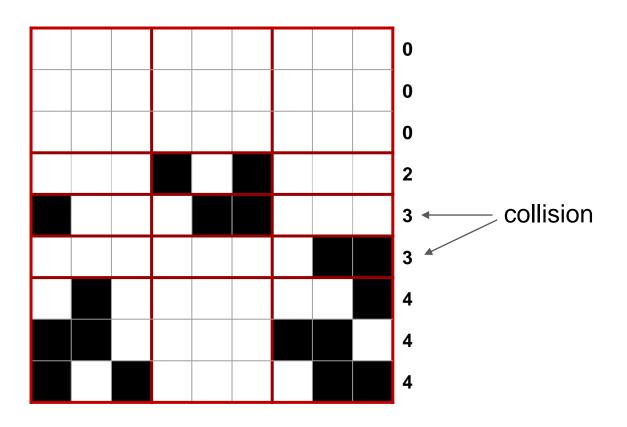
### The reordering algorithm - hashing the rows



### The reordering algorithm - hashing the rows

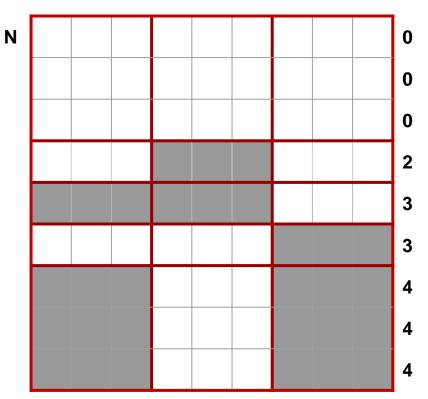


## The reordering algorithm – grouping by hash



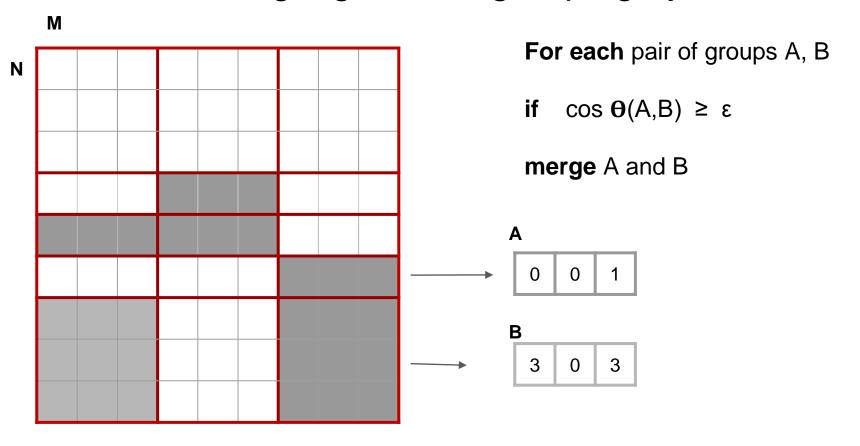
### The reordering algorithm - grouping by hash



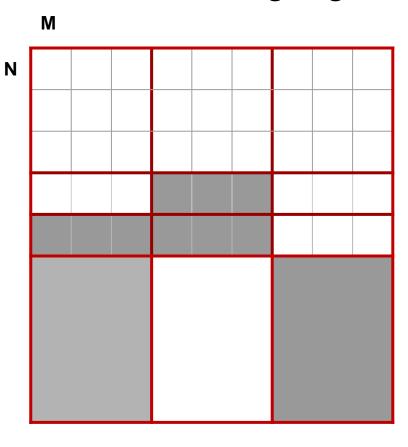


- 1) hashing: O(NxM) for CSR matrix with K entries: O(K)
- 1) sorting: O(N log N)
- 1) grouping: approx. NxM

### The reordering algorithm - grouping by cosine



### The reordering algorithm - grouping by cosine



For each pair of groups A, B

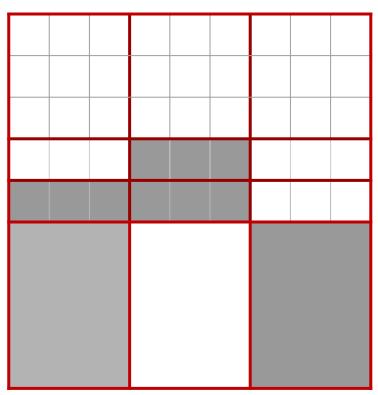
if  $\cos \Theta(A,B) \ge \varepsilon$ 

merge A and B

### The reordering algorithm - grouping by cosine

M

N



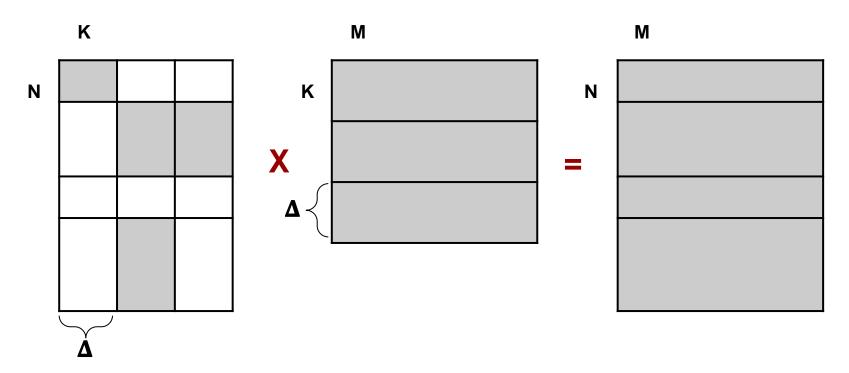
#### O(N\*N\*M)

But with some optimization:

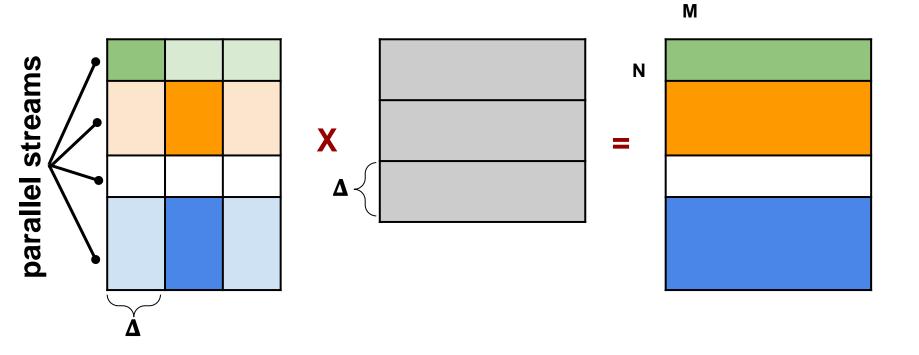
- Only groups are compared, not rows
- Only patterns are compared, not entries

Still, too expensive!

# Parallel multiplication

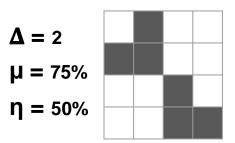


### Parallel multiplication

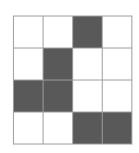


### Reorder - experiments with synthetic matrices

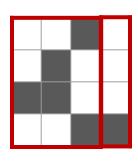
#### Synthetic block matrix



#### Scrambled



#### **Partitioned**



$$\Delta$$
' = 3

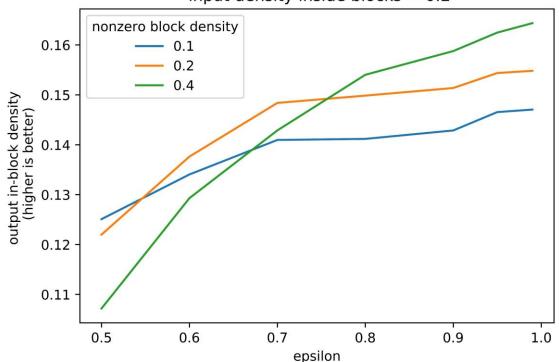
 $\Delta$  = original block size

 $\mu$  = original in-block density

 $\eta$  = original nonzero block density

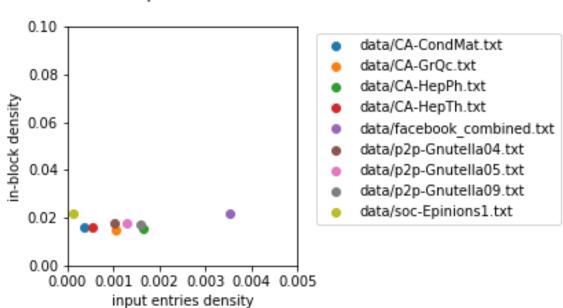
### Reorder - experiments with synthetic matrices

algorithm block size = 35 input block size = 64 input density inside blocks = 0.2



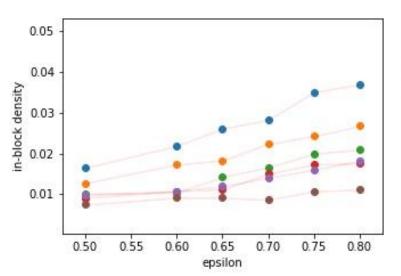
### Reorder - experiments with real matrices

algo\_block\_size:64 scramble:1 epsilon:0.6

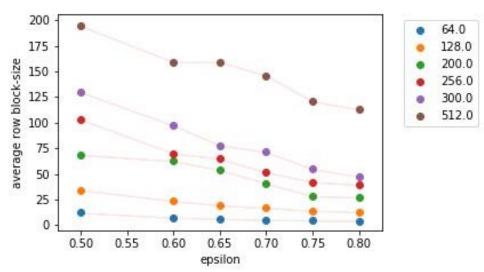


### Reorder - experiments with real matrices

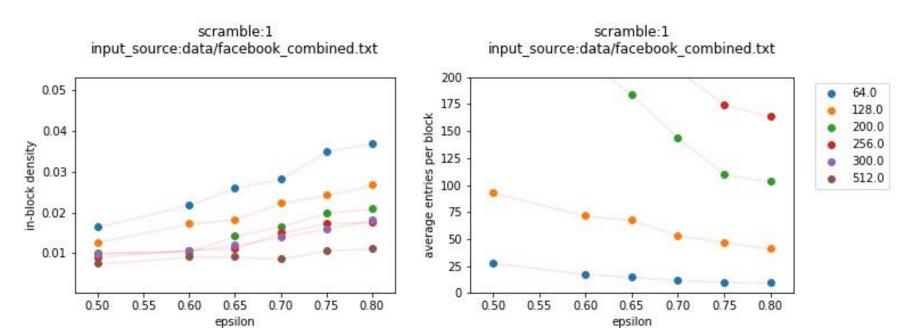
scramble:1 input\_source:data/facebook\_combined.txt



scramble:1 input source:data/facebook combined.txt



### Reorder - experiments with real matrices



scramble:1 input\_source:data/p2p-Gnutella04.txt

64.0

128.0

### Reorder

average row block-size 200.0 256.0 150 scramble:1 300.0 input source:data/p2p-Gnutella04.txt 512.0 100 0.030 50 0.025 0 in-block density 0.70 0.50 0.55 0.60 0.65 0.75 0.80 0.020 epsilon 0.015 200 64.0 175 128.0 0.010 average entries per block 200.0 150 256.0 0.005 125 300.0 512.0 0.000 100 0.50 0.55 0.60 0.65 0.70 0.75 0.80 75 epsilon 50 25

0

0.50

0.55

0.60

0.65

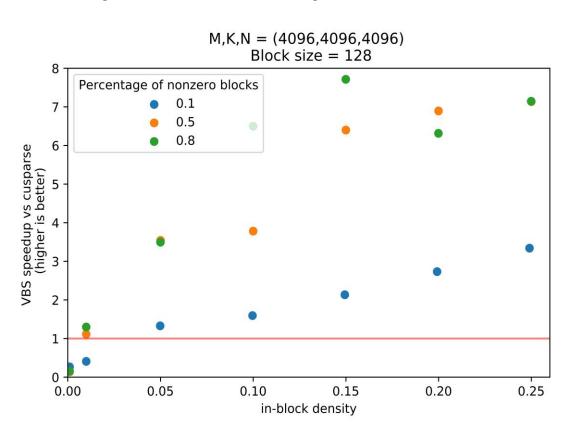
epsilon

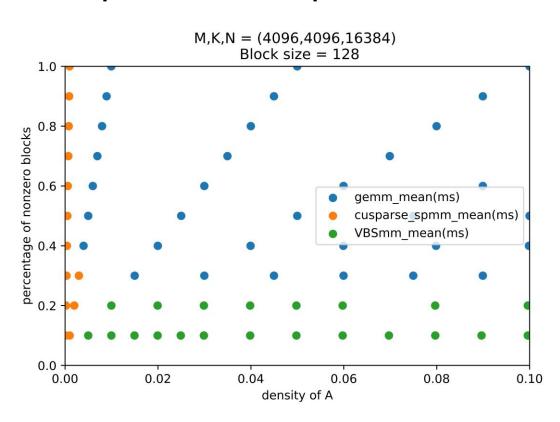
0.70

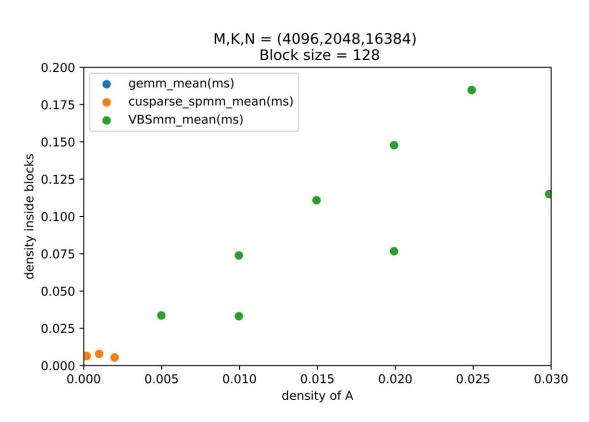
0.75

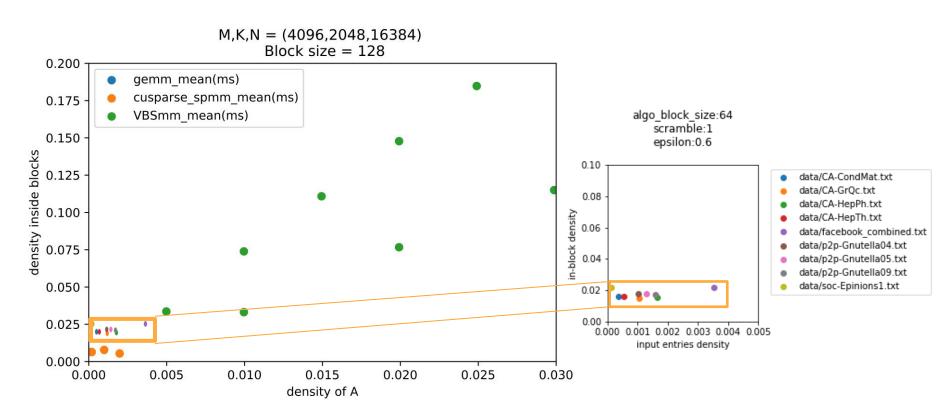
0.80

200









### Future directions – Better, faster reordering

- Parallel implementation
- Local Sensitive Hashing

### Future directions – LSH

Faster reordering algorithm through local sensitive hashing and nearest neighbours search

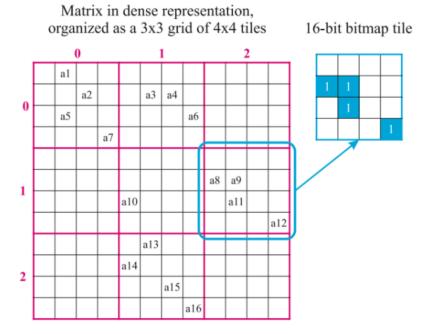
MinHash (random permutations): approximate Jaccard similarity

SimHash (random projections): approximate cosine similarity

### Future directions - Compare to related work

Accelerating sparse matrix–matrix multiplication with GPU Tensor Cores<sup>☆</sup>

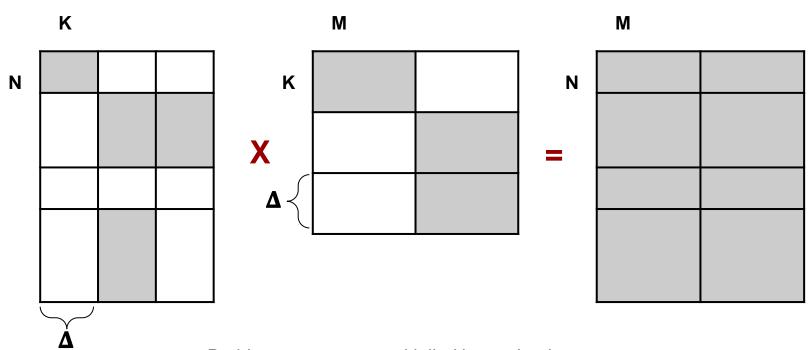
Orestis Zachariadis a,\*, Nitin Satpute a, Juan Gómez-Luna b, Joaquíi



### Future directions - Tensor-core implementation

- Optimal size of block partition?
- Best way to parallelize?
- Focused experiments

### Future directions - Sparse-Sparse multiplication



Problem: zeros gets multiplied by each other Number of unnecessary operations is squared