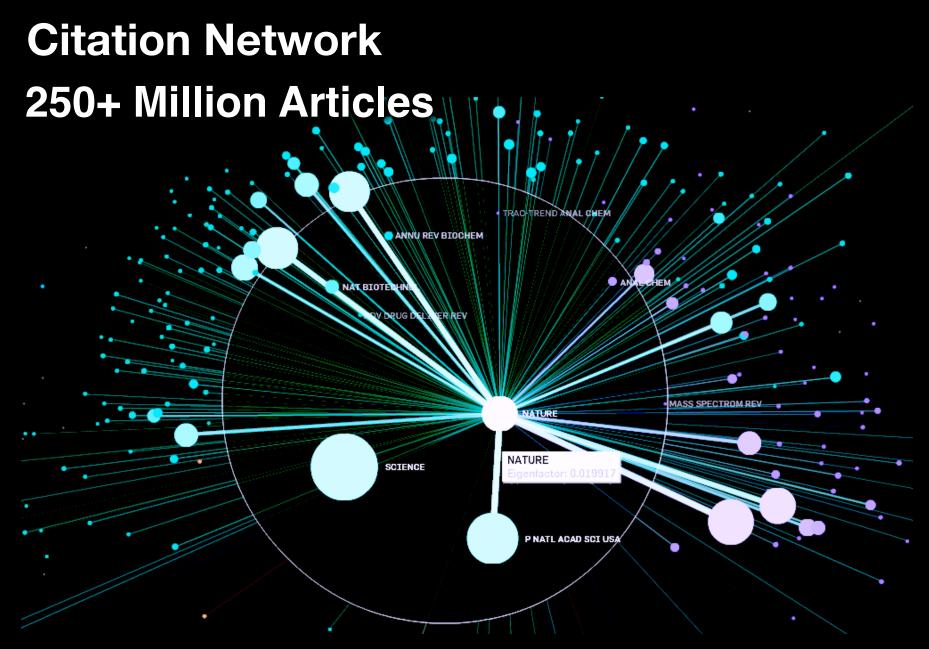
M-Flash:

Fast **Billion-Scale** Graph Computation Using a Bimodal Block Processing Model

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Internet 4+ Billion Web Pages



Many More



Who-follows-whom (310 million monthly active users)

Amazon Who-buys-what (300+ million users)

at&t cellphone network

Who-calls-whom (130+ million users)

Protein-protein interactions

200 million possible interactions in human genome

Sources: www.selectscience.net www.phonedog.com www.mediabistro.com www.practicalecommerce.com

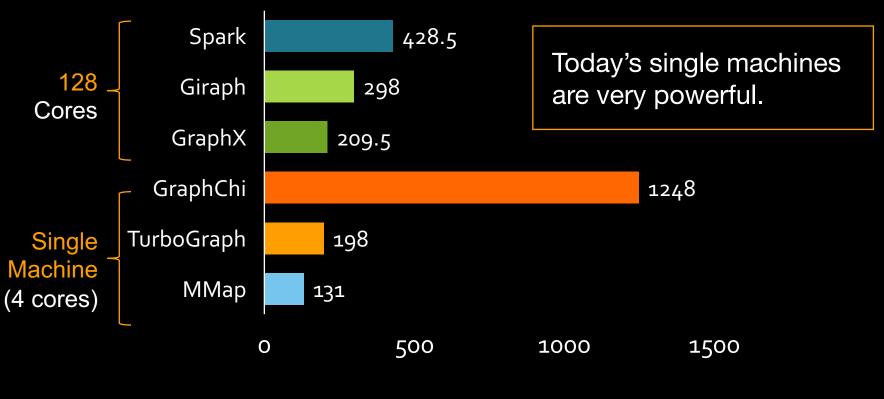
Large Graphs Are Common

Graph	Nodes	Edges
YahooWeb	1.4 Billion	6 Billion
Symantec Machine-File Graph	1 Billion	37 Billion
Twitter	104 Million	3.7 Billion
Phone call network	30 Million	260 Million

Takes Most Space

Scalable Graph Computation on Single Machines

PageRank Runtime (s) on Twitter Graph (1.5 billion edges; 10 iterations, lower is better)



Can we do even better?

McSherry, Frank, Michael Isard, and Derek G. Murray. "Scalability! But at what COST?." 15th Workshop on Hot Topics in Operating Systems (HotOS XV). 2015. Lin, Zhiyuan, et al. "Mmap: Fast billion-scale graph computation on a pc via memory mapping." Big Data (Big

Data), 2014 IEEE International Conference on. IEEE, 2014.

M-Flash:

Fast **Billion-Scale** Graph Computation Using a Bimodal Block Processing Model

Our Observation #1: I/O is Bottleneck

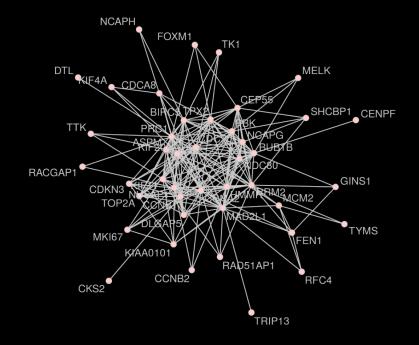
Graph edges need to be stored on disk. Symantec graph: 37 billion edges, 200+ GB

Disk access is much slower than RAM.

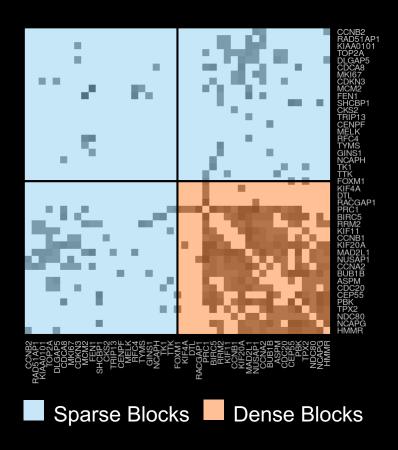
Goal: Reduce I/O, especially random accesses

Our Observation #2: Real-world graphs are sparse.

Adjacency matrix contains dense and sparse blocks



https://web.stanford.edu/class/bios221/labs/networks/lab_7_networks. html

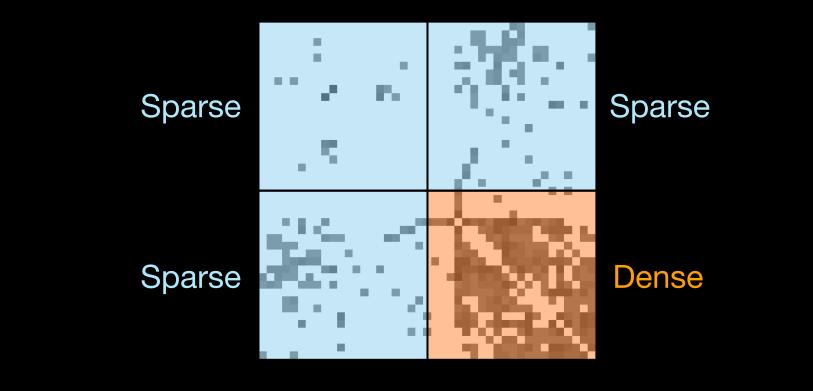


M-Flash's Solutions

 Determine edge block types (dense and sparse)

2. Design efficient processing approaches for each block type

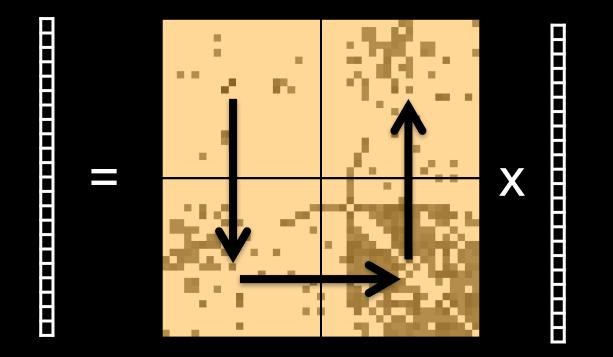
Determine Block Types In Pre-processing



BlockType =
$$\int$$
 Sparse, if $\frac{I/O \text{ cost if treated as Sparse}}{I/O \text{ cost if treated as Dense}} < 1$
Dense, otherwise

Dense Block Processing

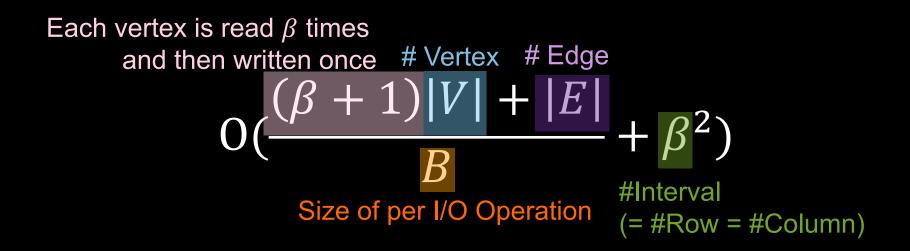
(Assuming all blocks are dense)

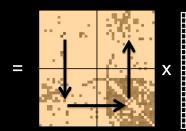


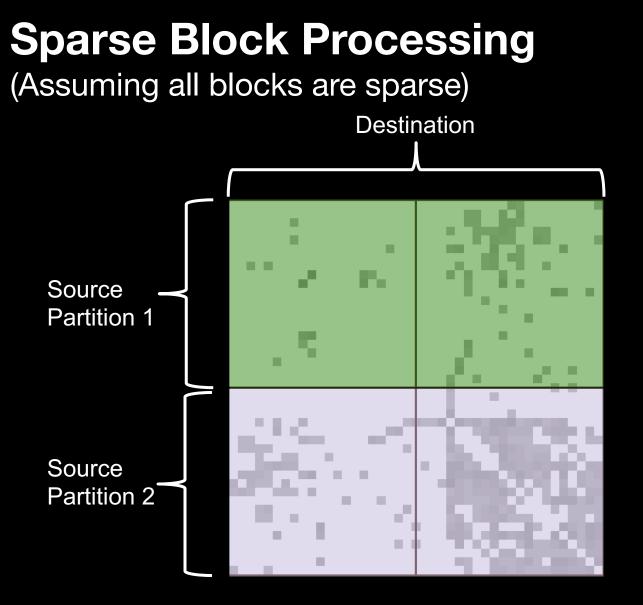
New vertex values

Old vertex values

I/O Cost for Dense Block Processing



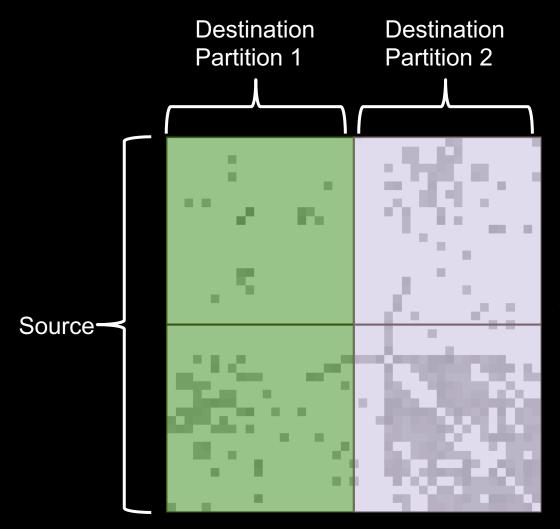




Source Partition: Sequential Read

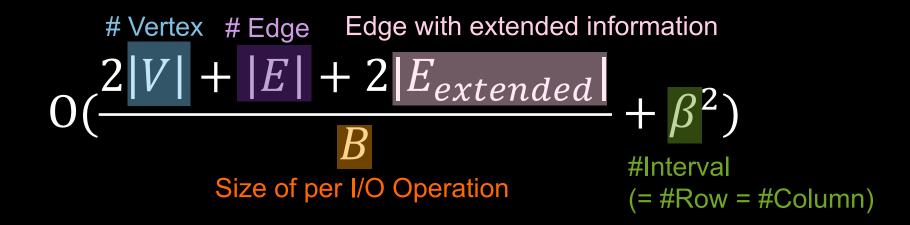
Sparse Block Processing

(Assuming all blocks are sparse)

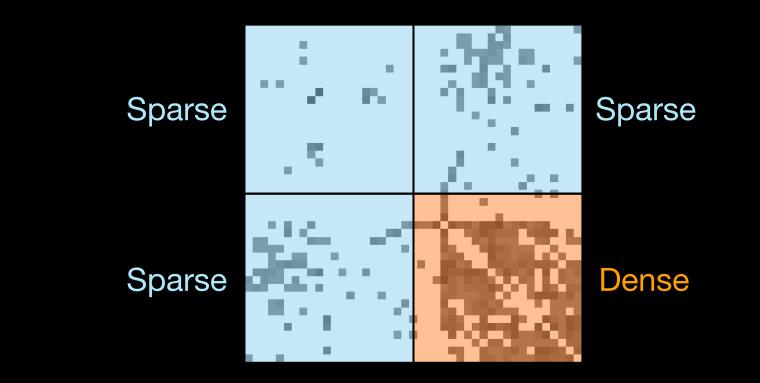


Destination Partition: Sequential Write

I/O Cost for Sparse Block Processing



Bimodal Block Processing



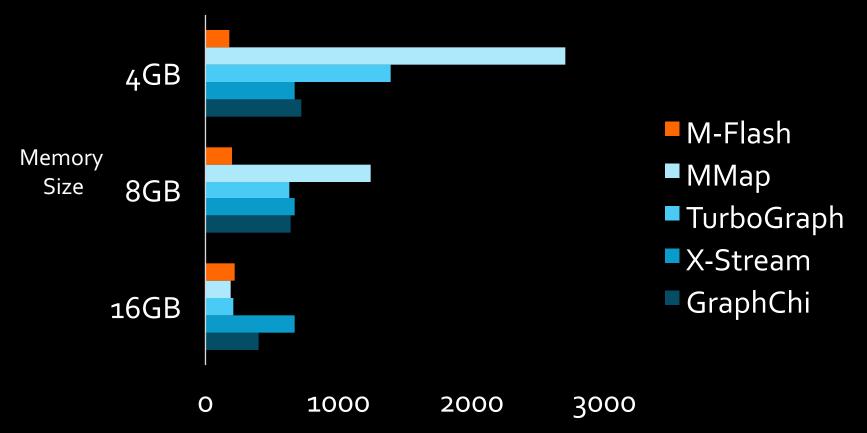
BlockType =
$$-\begin{cases} Sparse, if \frac{I/O \text{ cost if treated as Sparse}}{I/O \text{ cost if treated as Dense}} < 1 \\ Dense, otherwise \end{cases}$$

Large Graphs Used in Evaluation

Graph	Nodes	Edges
LiveJournal	5 Million	69 Million
Twitter	41 Million	1.5 Billion
YahooWeb	1.4 Billion	6.6 Billion
R-Mat (Synthetic)	4 Billion	12 Billion

Runtime of M-Flash

PageRank Runtime (s) on 6 billion edge YahooWeb Graph (1 iteration, shorter is better)







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Fast **Billion-Scale** Graph Computation Using a Bimodal Block Processing Model

- Fastest single-node graph computing framework
- Innovative bimodal design that addresses varying edge density in real-world graphs
- M-Flash Code: <u>https://github.com/M-Flash/m-flash-cpp</u>
- MMap Project: <u>http://poloclub.gatech.edu/mmap/</u>

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