



# AN EXPERIMENTAL ANALYSIS OF A COMPACT GRAPH REPRESENTATION

Endrias Kahssay

# WHY GRAPH COMPRESSION?

Graphs are getting larger (billions of edges) and take a large amount of memory to store.

A compression scheme can result in them fitting on RAM / cache.

Large performance penalty for an access to disk for graphs that don't fit.

# THE ORDER OF VERTICES MATTERS

Most graph algorithms access neighbors of a vertex  $u$  after accessing  $u$ .

We want a representation that puts edges of neighbors of vertices to be close to each other spatially.

In practice, this means making the labels of related vertices as close as possible.

# PAPER CONTRIBUTION

This paper empirically evaluates a representation that has both compression and good order.

Compression of edges using consecutive edge difference.

Order based on separator trees; a structure that attempts to encode a hierarchical relationship among nodes.

# SEPARATORS

An edge-separator is a set of edges (vertices) whose removal partitions the graph into two almost equal sized regions.

The min separator for a graph is one that minimizes the set of edges (vertices) to be removed.

We say that a graph has a good separator if its significantly better than a random graph.

# SEPERATORS IN REAL GRAPHS

Real world graphs have good separators because they are based on communities!

Include citation graphs, phone call graphs, friendship relationships, etc.

Useful for partitioning graphs for parallel processing, clustering, etc.

# COMPRESSED REPRESENTATION

Rep: an encoding where neighbors for each vertex are stored in a difference-encoded adjacency list.

I.e if a vertex has a sorted neighbor  $v_1, v_2, v_3, \dots$ , we store  $v_1 - v, v_2 - v_1$  contiguously in bits.

Difference stored using a logarithmic code which uses  $O(\log d)$  bits to encode a diff of size  $d$ .

# REPRESENTATION

Adjacency list concatenated to form an adjacency table.

An indexing structure is used to access the adjacency list for vertices, which is also compressed.

Not theoretically optimal, but works well in practice.

# REPRESENTATION

- 1) Generate edge separator tree for the graph.
- 2) Label in the vertices in-order across the leaves.
- 3) Use an adjacency table to represent labeled graph.

In previous work, the authors describe an  $O(n)$  bit encoding with  $O(1)$  access time for graphs satisfying  $O(n^c)$  for  $c < 1$  separator bound.

# IMPLEMENTATION: SEPARATOR TREES

Collapses edges until a single vertex remains.

Many heuristics to collapse edges, paper used  $w(E_{AB}) / s(A)s(B)$ .

$w(E_{AB})$  is the number of edges between A and B, and  $S(B)$  is the number of vertices in the multi vertex B.

# INDEXING STRUCTURE

Used to map a vertex to its start location.

Tradeoff between space and look speed.

Uses semi-direct-16, which stores the start location for 16 vertices in 5 32-bit words.

Based on storing offsets from some set of vertices encoded, and if an offset doesn't fit, a pointer is stored to it instead.

# CODES AND DECODING

Originally used gamma code, which represents an integer  $d$  using a unary code for  $\lceil \log d \rceil$  followed by binary code for  $d - 2^{\lceil \log d \rceil}$ .

Decoding gamma code is expensive (even though  $O(1)$ ).

Developed snip, nibble, and byte code, optimized for machine word access based on continue bits and chunking.

# DYNAMIC VERSION

In dynamic version, the edge list of each vertex is dynamically managed, so can't be stored contiguously.

Initially array with one memory block for each vertex.

If additional memory is needed, vertex gets additional blocks from a pool.

Blocks connected in a linked list.

# DYNAMIC VERSION

Hashing technique to reduce size of pointers to 8 bits.

A separate pool for each 1024 vertices for locality.

Caching to avoid repetitively encoding and decoding neighbors.

# EXPERIMENTS

Test graphs include 3d mesh graphs, street connectivity graphs, webpage connectivity from Google, programming contest, and circuits.

| Graph   | Vtxs    | Edges   | Max Degree | Source          |
|---------|---------|---------|------------|-----------------|
| auto    | 448695  | 6629222 | 37         | 3D mesh [35]    |
| feocean | 143437  | 819186  | 6          | 3D mesh [35]    |
| m14b    | 214765  | 3358036 | 40         | 3D mesh [35]    |
| ibm17   | 185495  | 4471432 | 150        | circuit [1]     |
| ibm18   | 210613  | 4443720 | 173        | circuit [1]     |
| CA      | 1971281 | 5533214 | 12         | street map [34] |
| PA      | 1090920 | 3083796 | 9          | street map [34] |
| googleI | 916428  | 5105039 | 6326       | web links [10]  |
| googleO | 916428  | 5105039 | 456        | web links [10]  |
| lucent  | 112969  | 363278  | 423        | routers [25]    |
| scan    | 228298  | 640336  | 1937       | routers [25]    |

# MACHINE SETUP

Two machines, each with 32-bit processors.

A .7GHz Pentium III processor with .1Ghz bus and 1GB of ram (cache line 32 bytes).

2.4Ghz, Pentium 4, with 4 processors, .8GHz bus, and 1GB of ram (cache line 128 bytes). Supports quad vectorization, and hardware prefetching.

# BENCHMARK

Benchmark on depth-first-search, and also varying edge insertion order.

DFS visits in a non trivial order.

Use a character array to mark visited vertices.

# INSERTING EDGES

Consider three ways of inserting edges:

Linear: Insert all out edges for first vertex, then second, and so on.

Transpose: insert all in edge for first vertex, etc.

Random: random order of edges.

# PERFORMANCE FOR DFS, STATIC

| Graph      | Array  |         |        | Our Structure |        |         |       |         |       |         |       |          |        |
|------------|--------|---------|--------|---------------|--------|---------|-------|---------|-------|---------|-------|----------|--------|
|            | Rand   | Sep     |        | Byte          |        | Nibble  |       | Snip    |       | Gamma   |       | DiffByte |        |
|            | $T_1$  | $T/T_1$ | Space  | $T/T_1$       | Space  | $T/T_1$ | Space | $T/T_1$ | Space | $T/T_1$ | Space | $T/T_1$  | Space  |
| auto       | 0.268s | 0.313   | 34.17  | 0.294         | 10.25  | 0.585   | 7.42  | 0.776   | 6.99  | 1.063   | 7.18  | 0.399    | 12.33  |
| feocean    | 0.048s | 0.312   | 37.60  | 0.312         | 12.79  | 0.604   | 10.86 | 0.791   | 11.12 | 1.0     | 11.97 | 0.374    | 13.28  |
| m14b       | 0.103s | 0.388   | 34.05  | 0.349         | 10.01  | 0.728   | 7.10  | 0.970   | 6.55  | 1.320   | 6.68  | 0.504    | 11.97  |
| ibm17      | 0.095s | 0.536   | 33.33  | 0.536         | 10.19  | 1.115   | 7.72  | 1.400   | 7.58  | 1.968   | 7.70  | 0.747    | 12.85  |
| ibm18      | 0.113s | 0.398   | 33.52  | 0.442         | 10.24  | 0.867   | 7.53  | 1.070   | 7.18  | 1.469   | 7.17  | 0.548    | 12.16  |
| CA         | 0.920s | 0.126   | 43.40  | 0.146         | 14.77  | 0.243   | 10.65 | 0.293   | 10.55 | 0.333   | 11.25 | 0.167    | 14.81  |
| PA         | 0.487s | 0.137   | 43.32  | 0.156         | 14.76  | 0.258   | 10.65 | 0.310   | 10.60 | 0.355   | 11.28 | 0.178    | 14.80  |
| lucent     | 0.030s | 0.266   | 41.95  | 0.3           | 14.53  | 0.5     | 11.05 | 0.566   | 10.79 | 0.700   | 11.48 | 0.333    | 14.96  |
| scan       | 0.067s | 0.208   | 43.41  | 0.253         | 15.46  | 0.402   | 11.84 | 0.477   | 11.61 | 0.552   | 12.14 | 0.298    | 16.46  |
| googleI    | 0.367s | 0.226   | 37.74  | 0.258         | 11.93  | 0.405   | 8.39  | 0.452   | 7.37  | 0.539   | 7.19  | 0.302    | 13.39  |
| googleO    | 0.363s | 0.250   | 37.74  | 0.278         | 12.59  | 0.460   | 9.72  | 0.556   | 9.43  | 0.702   | 9.63  | 0.327    | 13.28  |
| <b>Avg</b> |        | 0.287   | 38.202 | 0.302         | 12.501 | 0.561   | 9.357 | 0.696   | 9.07  | 0.909   | 9.424 | 0.380    | 13.662 |

Table 2: Performance of our **static** algorithms compared to performance of an adjacency array representation. Space is in bits per edge; time is for a DFS, normalized to the first column, which is given in seconds.

separator-based representation with byte codes is a factor of

3.3 faster than adjacency arrays with random ordering but about 5% slower for the separator ordering.

6.6 more compact than adjacency list.

| Graph      | Linked List      |                  |                |                 |                  |                |        | Our Structure |                 |        |               |                 |        |
|------------|------------------|------------------|----------------|-----------------|------------------|----------------|--------|---------------|-----------------|--------|---------------|-----------------|--------|
|            | Random Vtx Order |                  |                | Sep Vtx Order   |                  |                | Space  | Space Opt     |                 |        | Time Opt      |                 |        |
|            | Rand<br>$T_1$    | Trans<br>$T/T_1$ | Lin<br>$T/T_1$ | Rand<br>$T/T_1$ | Trans<br>$T/T_1$ | Lin<br>$T/T_1$ |        | Block<br>Size | Time<br>$T/T_1$ | Space  | Block<br>Size | Time<br>$T/T_1$ | Space  |
| auto       | 1.160s           | 0.512            | 0.260          | 0.862           | 0.196            | 0.093          | 68.33  | 16            | 0.148           | 9.35   | 20            | 0.087           | 13.31  |
| feocean    | 0.136s           | 0.617            | 0.389          | 0.801           | 0.176            | 0.147          | 75.21  | 8             | 0.227           | 12.97  | 10            | 0.117           | 14.71  |
| m14b       | 0.565s           | 0.442            | 0.215          | 0.884           | 0.184            | 0.090          | 68.09  | 16            | 0.143           | 8.92   | 20            | 0.086           | 13.53  |
| ibm17      | 0.735s           | 0.571            | 0.152          | 0.904           | 0.357            | 0.091          | 66.66  | 12            | 0.205           | 10.53  | 20            | 0.118           | 14.52  |
| ibm18      | 0.730s           | 0.524            | 0.179          | 0.890           | 0.276            | 0.080          | 67.03  | 10            | 0.190           | 10.13  | 20            | 0.108           | 14.97  |
| CA         | 1.240s           | 0.770            | 0.705          | 0.616           | 0.107            | 0.101          | 86.80  | 3             | 0.170           | 10.62  | 5             | 0.108           | 15.65  |
| PA         | 0.660s           | 0.780            | 0.701          | 0.625           | 0.112            | 0.109          | 86.64  | 3             | 0.180           | 10.69  | 5             | 0.115           | 15.64  |
| lucent     | 0.063s           | 0.634            | 0.492          | 0.730           | 0.190            | 0.142          | 83.90  | 3             | 0.285           | 13.67  | 6             | 0.174           | 20.49  |
| scan       | 0.117s           | 0.735            | 0.555          | 0.700           | 0.188            | 0.128          | 86.82  | 3             | 0.290           | 15.23  | 8             | 0.170           | 28.19  |
| googleI    | 0.975s           | 0.615            | 0.376          | 0.774           | 0.164            | 0.096          | 75.49  | 4             | 0.211           | 12.04  | 16            | 0.125           | 28.78  |
| googleO    | 0.960s           | 0.651            | 0.398          | 0.786           | 0.162            | 0.108          | 75.49  | 5             | 0.231           | 13.54  | 16            | 0.123           | 26.61  |
| <b>Avg</b> |                  | 0.623            | 0.402          | 0.779           | 0.192            | 0.108          | 76.405 |               | 0.207           | 11.608 |               | 0.121           | 18.763 |

Table 4: The performance of our **dynamic** algorithms compared to linked lists. For each graph we give the space- and time-optimal block size. Space is in bits per edge; time is for a DFS, normalized to the first column, which is given in seconds.

Label ordering has up to a 7x effect Insertion order has up to a factor of 11x effect.

# ALGORITHMIC BENCHMARKS

PageRank (summing neighbors), and Bipartite matching (based on DFS).

| Representation | Time (sec) |       | Space<br>(b/e) |
|----------------|------------|-------|----------------|
|                | PIII       | P4    |                |
| Dyn-B4         | 30.40      | 11.05 | 17.54          |
| Dyn-N4         | 32.96      | 12.48 | 13.28          |
| Dyn-B8         | 26.55      | 9.23  | 19.04          |
| Dyn-N8         | 30.29      | 11.25 | 15.65          |
| Gamma          | 38.56      | 15.60 | 9.63           |
| Snip           | 34.19      | 13.38 | 9.43           |
| Nibble         | 26.38      | 10.94 | 9.72           |
| Byte           | 21.09      | 8.04  | 12.59          |
| ArrayOrdr      | 21.12      | 6.38  | 37.74          |
| ArrayRand      | 33.83      | 27.59 | 37.74          |
| ListOrdr       | 30.96      | 6.12  | 75.49          |
| ListRand       | 44.56      | 28.33 | 75.49          |

# CONCLUSION

Simple and fast separator heuristic works well in practice, and real world graphs have small separators.

Decoding cost is small even for simple graph algorithms like DFS.

Order of vertices makes a big difference in terms of a performance (up to 11x).

Separator works well for other graph representations.

# STRENGTHS AND WEAKNESS

Paper explains background information well.

Rigorous experiment section, and admits limitations in comparisons.

Would have been interesting to test other traversal patterns like BFS.

# QUESTIONS

What would these results look like if they were tested on a modern machine/ with NUMA?

How well would separators work for laying the graph in a NUMA memory setting?