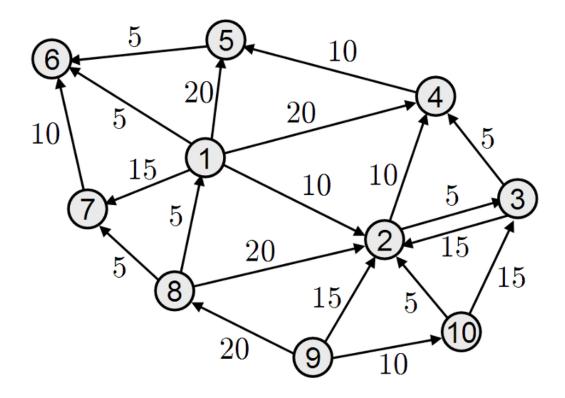
R244 Michael Chi Ian Tang

### Green-Marl A DSL for Easy and Efficient Graph Analysis S. Hong, H. Chafi, E. Sedlar, K.Olukotun

# Background

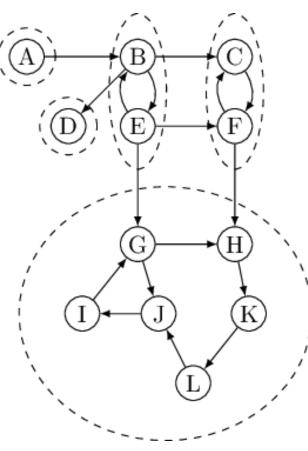
### Background

• Graph Analysis - Extract information from a graph dataset



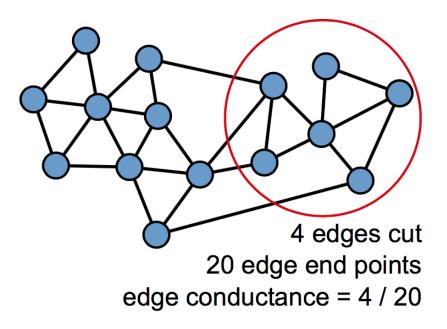
### Background – Graph Analysis

Strongly Connected Components

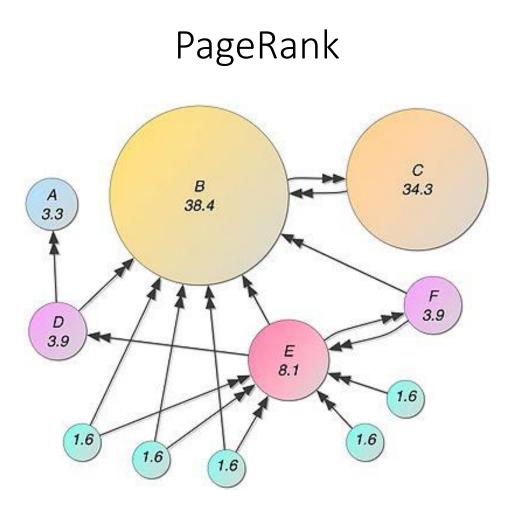


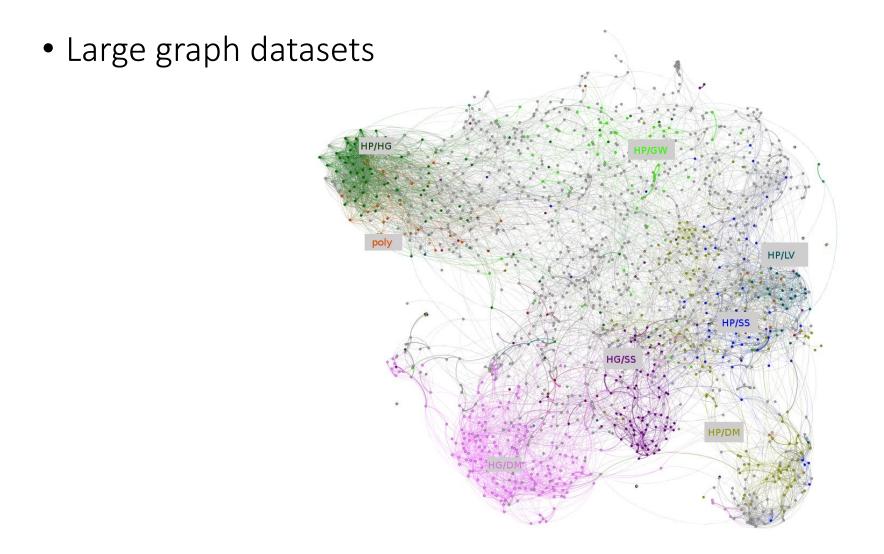
### Background – Graph Analysis

#### Conductance



### Background – Graph Analysis





}

• Intuitive Implementation vs Capturing Parallelism

```
BFS(G, s) {
    initialize vertices;
    Q = {s};
    while (Q not empty) {
        u = RemoveTop(Q);
        for each v ∈ u->adj {
            if (v->color == WHITE)
               v->color = GREY;
               v->d = u->d + 1;
               v->p = u;
               Enqueue(Q, v);
        }
        u->color = BLACK;
    }
```

-	gorithm 1. CUDA_BFS (Graph $G(V, E)$ , Source Vertex S)
1:	Create vertex array $V_a$ from all vertices and edge Array $E_a$ from all edges in $G(V, E)$ ,
2:	Create frontier array $F_a$ , visited array $X_a$ and cost array $C_a$ of size V.
3:	Initialize $F_a$ , $X_a$ to false and $C_a$ to $\infty$
4:	$F_a[S] \leftarrow \text{true}, C_a[S] \leftarrow 0$
5:	while <i>F<sub>a</sub></i> not Empty do
6:	for each vertex V in parallel do
7:	Invoke CUDA_BFS_KERNEL( $V_a, E_a, F_a, X_a, C_a$ ) on the grid.
8:	end for
9:	end while

• Challenges - Capacity, Performance, Implementation



- DSL (Domain-specific language)
  - Separation of programming and optimization
  - Intuitive implementation of graph algorithms
  - Exposes data-level parallelism

```
/_____
// Computing Conductance
//------
Procedure conductance(G: Graph, member: N P<Int>(G), num: Int) : Double {
 Int Din, Dout, Cross;
 Din = Sum(u:G.Nodes)(u.member==num){u.Degree()};
                                                      // Compute degree sum of inside nodes.
 Dout = Sum(u:G.Nodes) (u.member!=num) {u.Degree()};
                                                       // Compute degree sum of outside nodes.
                                                       // Count number of crossing edges.
 Cross = Sum(u:G.Nodes) (u.member==num) {
         Count(j:u.Nbrs)(j.member!=num);
                                                       // (Count is a syntactic sugar to Sum(..) {1}
 Double m = (Din < Dout) ? Din : Dout;
 If (m ==0) Return (Cross == 0) ? 0.0 : +INF;
 Else Return (Cross / m);
```

- Simple language constructs
  - Primitive types: Bool, Int, Long, Float, Double
  - Graph types: Undirected, Directed
  - Types bounded to graphs: Node, Edge
  - Collection types: Set, Order, Sequence
  - Traversal Schemes: BFS, DFS

```
Procedure foo(G1, G2:Graph, n:Node(G1)) {
   Node(G2) n2; // a node of graph G2
   n2 = n; // type-error (bound to different graphs)
   Node_Prop<Int>(G1) A; //integer node property for G1
   n.A = 0;
   Node_Set(G1) S; // a node set of G1
   S.Add(n);
```

• Deferred assignment and Reductions

```
Foreach(s:G.Nodes) {
    // no conflict. t.X gives 'old' value
    s.X <= Sum(t:s.Nbrs) {t.X} @ s
}
// All the writes to X becomes visible simultaneously
// at the end of the s iteration.</pre>
```

## **Optimizations & Compilation**

### Optimizations

- Architecture independent optimizations
  - Loop Fusion, Hoisting Definitions, Reduction Bound Relaxation, Flipping Edges

```
133 Foreach(t:G.Nodes)(f(t))
134 Foreach(s:t.InNbrs)(g(s))
135 t.A += s.B;
136 Foreach(s:G.Nodes)(g(s))
137 Foreach(t:s.OutNbrs)(f(t))
138 t.A += s.B;
```

Example of Flipping Edges Optimization

### Optimizations

- Architecture dependent optimizations
  - Selection of Parallel Regions, Deferred Assignment and Saving BFS Children, Set-Graph Loop Fusion

139	<b>Node_Set</b> S(G); //
140	Foreach(s: S.Items)
141	s.A = x(s.B);
142	<pre>Foreach(t: G.Nodes)(g(t))</pre>
143	t.B = y(t.A)

#### becomes

Example of Set-Graph Loop Fusion Optimization

### Compilation

• Into general-purpose languages, e.g. C++ (using graph library)

222 Foreach(s:G.Nodes)
223 For(t: s.Nbrs)
224 s.A = s.A + t.B;

#### becomes

## Experiments & Comparisons

### **Concise Representation**

• Fewer lines-of-code (LOC) for many problems

	LOC	LOC	
Name	Original	Green-Marl	Source
BC	350	24	[9] (C OpenMp)
Conductance	42	10	[9] (C OpenMp)
Vetex Cover	71	25	[9] (C OpenMp)
PageRank	58	15	[2] (C++, sequential)
SCC(Kosaraju)	80	15	[3] (Java, sequential)

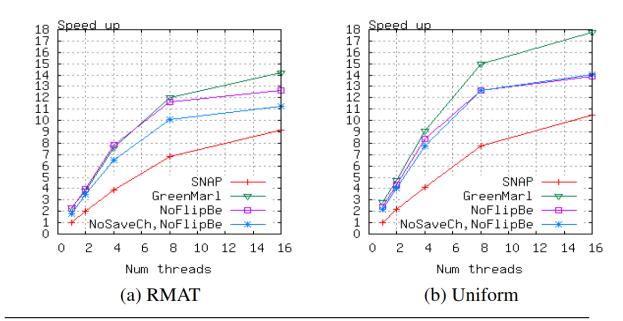
**Table 3.** Graph algorithms used in the experiments and their Linesof-Code(LOC) when implemented in Green-Marl and in a general purpose language.

### Experiments

- Betweenness Centrality
  - Compared to SNAP library

```
Procedure Compute_BC(
     G: Graph, BC: Node_Prop<Float>(G)) {
3
      G.BC = 0;
                       // initialize BC
      Foreach(s: G.Nodes)
       // define temporary properties
6
        Node_Prop<Float>(G) Sigma;
        Node_Prop<Float>(G) Delta;
8
        s.Sigma = 1; // Initialize Sigma for root
9
       // Traverse graph in BFS-order from s
10
        InBFS(v: G.Nodes From s)(v!=s) {
11
          // sum over BFS-parents
12
          v.Sigma = Sum(w: v.UpNbrs) {w.Sigma};
13
14
       // Traverse graph in reverse BFS-order
15
        InRBFS (v!=s)
          // sum over BFS-children
16
17
          v.Delta = Sum (w:v.DownNbrs)
18
             v.Sigma / w.Sigma * (1+ w.Delta)
19
          };
20
          v.BC += v.Delta @s; //accumulate BC
21
```

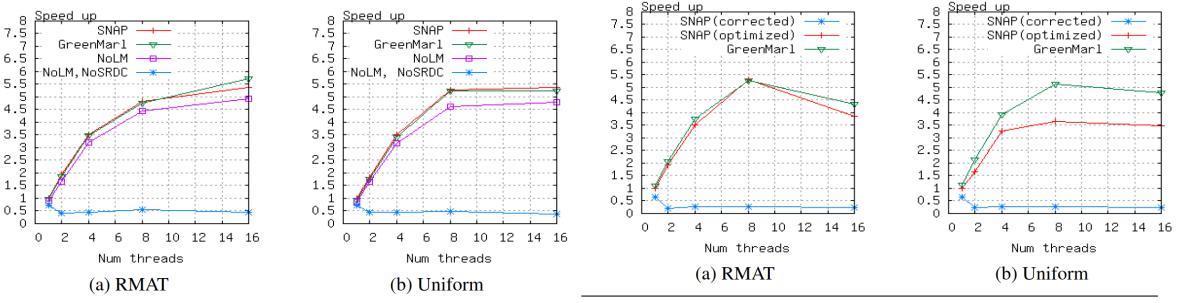
Figure 1. Betweenness Centrality algorithm described in Green-Marl



**Figure 4.** Speed-up of Betweenness Centrality. Speed-up is over the SNAP library [9] version running on a single-thread. NoFlipBE and NoSaveCh means disabling the *Flipping Edges* (Section 3.3) and *Saving BFS Children* (Section 3.5) optimizations respectively.

### Experiments

Conductance



• Vertex Cover

**Figure 5.** Speed-up of Conductance. Speed-up is over the SNAP library [9] version running on a single-thread. NoLM and NoSRDC means disabling the *Loop Fusion* (Section 3.3) and *Reduction on Scalars* (Section 3.5) optimizations, respectively.

**Figure 6.** Speed-up of Vertex Cover implemented in Green-Marl and two versions of the corrected SNAP implementation SNAP which had a data-race. The first version, SNAP(correct) utilizes a simple locking approach. The second version, SNAP(optimized), uses a more advanced test and test-and-set scheme. A small instance (100k nodes, 800k edges) was used in this experiment.

# Critique

- 1. Intuitive, concise implementation of algorithms
- 2. Transparent, automatic optimizations through compilation
- 3. Wider range of optimizations using domain-specific knowledge
- 4. Architecture-dependent optimizations
- 5. High architecture portability
- 6. Easy integration into current workflow

### Criticism

- 1. Limited to graphs which fits into RAM
- 2. Backend optimized for CPU execution only
- 3. Limited comparison with related works

### Conclusion

- A domain-specific language that is
  - Portable
  - Concise and intuitive
  - Efficient
  - Easy to integrate into workflow
- Require more work on
  - Scalability
  - Performance evaluation