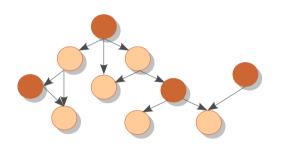
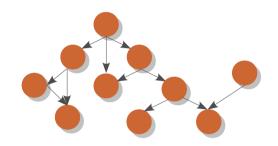
Data-driven versus Topology-driven Irregular Computations on GPUs





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What is IrReg_uL_ari^Ty?

 Data-access or control patterns are unpredictable at compile time.

Irregular data-access

```
int a[N], b[N], c[N];
readinput(a);
c[5] = b[a[4]];
```

Irregular control-flow

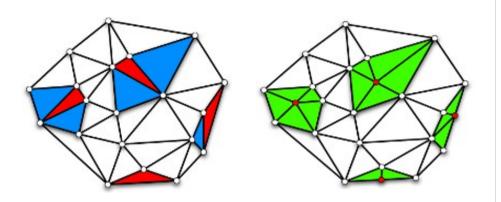
```
int a[N];
readinput(a);

if (a[4] > 30) {
    ...
}
```

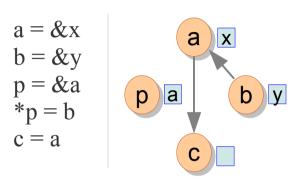
Needs dynamic techniques

Pointer-based data structures often contribute to irregularity.

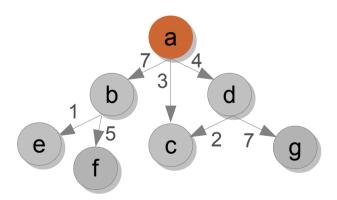
Examples of Irregular Programs



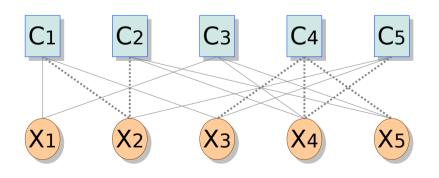
Delaunay Mesh Refinement



Points-to Analysis

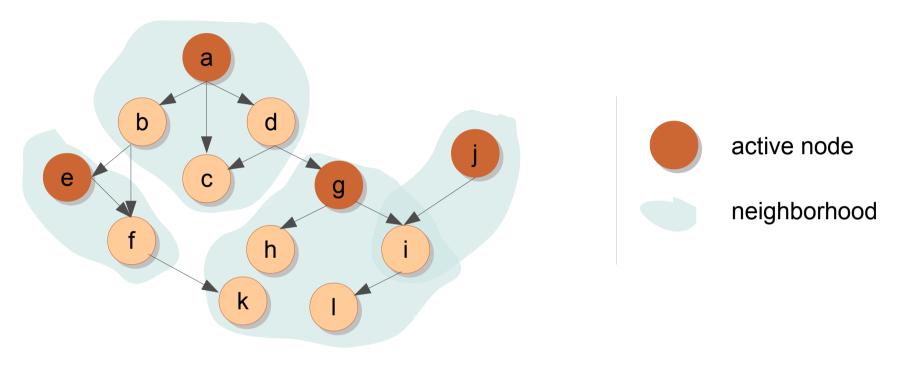


Single-source shortest paths



Survey Propagation

Operator Formulation



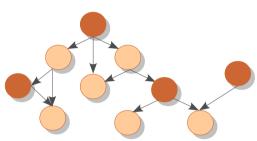
- Computation is modeled as an iterated application of an operator
- Data-driven processing: operator is applied only at the nodes where there might be work to be done
- Topology-driven processing: operator is applied at all the nodes even if there is no work to do at some nodes

Focus of This Work

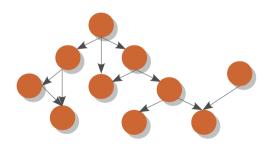
- Study and optimize topology-driven and data-driven irregular computations on GPUs
 - On GPUs, irregular algorithms are usually implemented in a topology-driven manner
 - On multi-core CPUs, irregular algorithms are usually implemented in a data-driven manner



- Data-driven versions usually perform better
- Topology-driven versions perform better if additional algorithmic properties can be exploited
- A hybrid approach can outperform the two



data-driven



topology-driven

Outline

- Motivation
- → Data-driven approach
- Topology-driven approach
- Hybrid approach
- Evaluation
- Conclusions

Outline

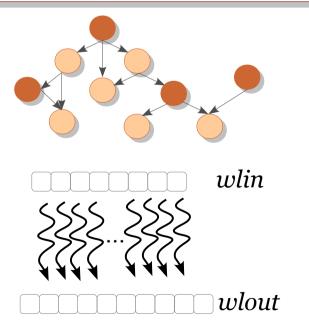
- Motivation
- → Data-driven approach
 - dual-worklist
 - hierarchical worklists
 - work-chunking
 - atomic-free worklist update
 - work-donation
 - variable kernel configuration
- Topology-driven approach
- Hybrid approach
- Evaluation
- Conclusions

Data-driven: Base Version

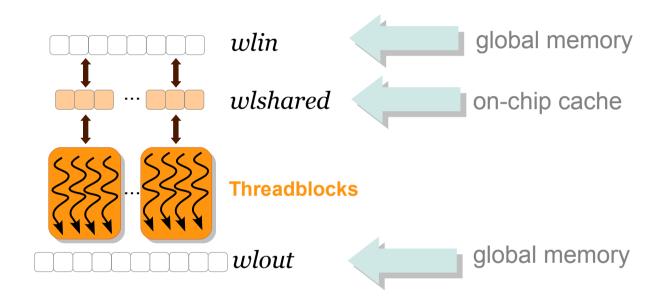
```
cpu qpu
main {
    read input
    transfer input
    initialize kernel
    initialize_worklist(wlin)
    clear wlout
    while wlin not empty {
         operator(wlin, wlout, ...)
         transfer wlout size
         clear wlin
         swap(wlin, wlout)
    transfer results
```

```
sssp_operator(wlin, wlout, ...) {
    src = wlin[...]
    dsrc = distance[src]
    forall edges (src, dst, wt) {
        ddst = distance[dst]
        altdist = dsrc + wt

    if altdist < ddst
        distance[dst] = altdist
        wlout.push(dst)
} }</pre>
```

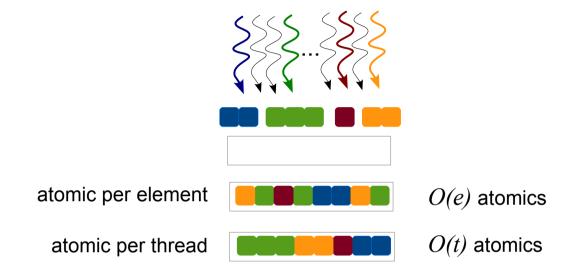


Data-driven: Hierarchical Worklist



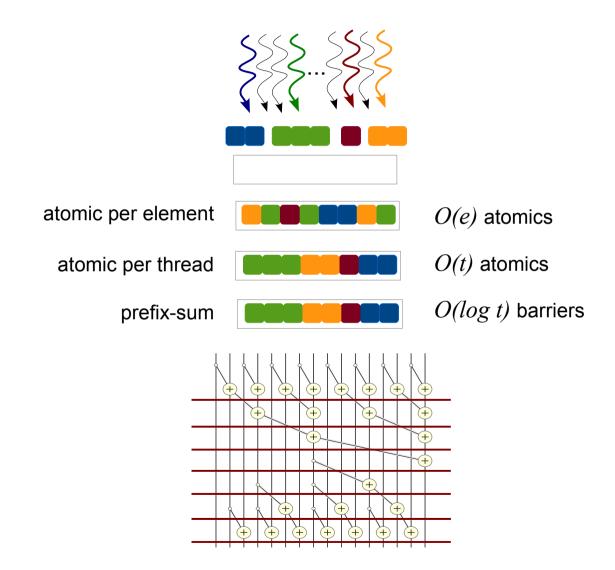
- Worklist exploits memory hierarchy
- Makes judicious use of limited on-chip cache

Data-driven: Work Chunking



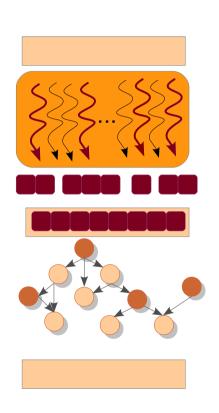
- Reserves space for multiple work-items in a single atomic
- May reduce overall synchronization

Data-driven: Atomic-free Worklist Update



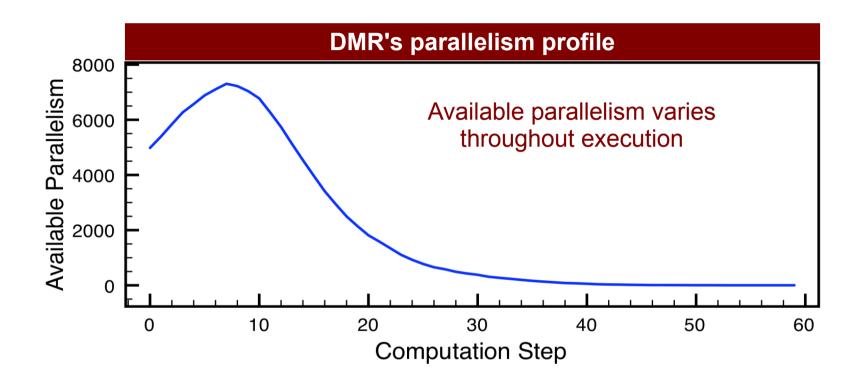
Data-driven: Work Donation

```
donate_kernel {
     shared donationbox[...];
    // determine if I should donate
    --barrier--
    // donate
    --barrier--
    // operator execution
    // empty donation box
```



Work-donation improves load balance

Data-driven: Variable Kernel Configuration



- Varying configuration improves work-efficiency
- It also reduces conflicts and may improve performance

Outline

- Motivation
- Data-driven approach
- Topology-driven approach
 - no global worklists
 - kernel unrolling
 - exploiting memory hierarchy
 - improved memory layout
- Hybrid approach
- Evaluation
- Conclusions

Topology-driven: Base Version

```
main {
    read input
    transfer input
    initialize_kernel
    do {
        transfer false to changed
        operator(...)
        transfer changed
    } while changed
    transfer results
```

Topology-driven: Kernel Unrolling

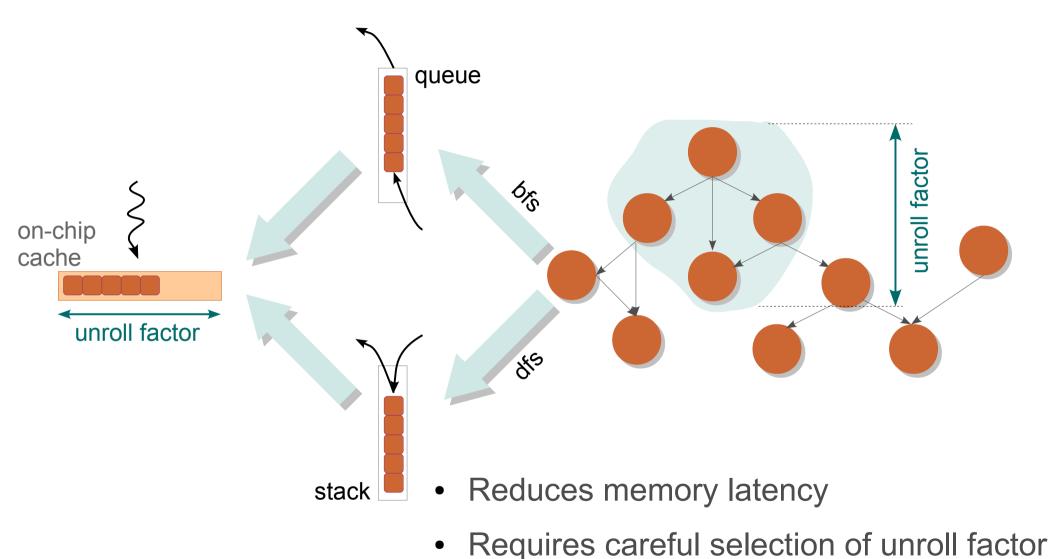
```
sssp_operator(src) {
    dsrc = distance[src]

forall edges (src, dst, wt) {
    ddst = distance[dst]
    altdist = dsrc + wt

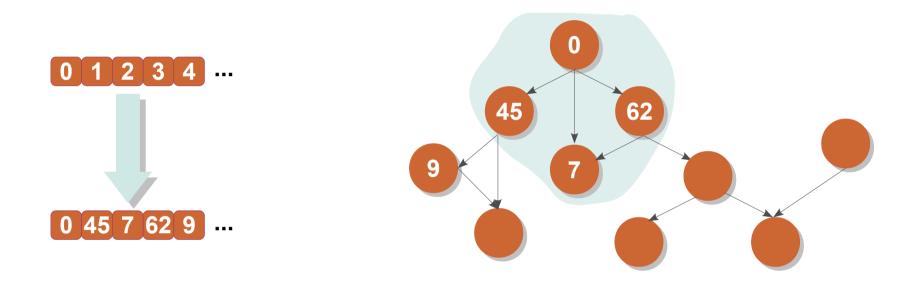
    if altdist < ddst
        distance[dst] = altdist
    }
}
```

- Improves amount of computation per thread invocation
- Need to ensure absence of races
- Propagates information faster

Topology-driven: Exploiting Memory Hierarchy



Topology-driven: Improved Memory Layout

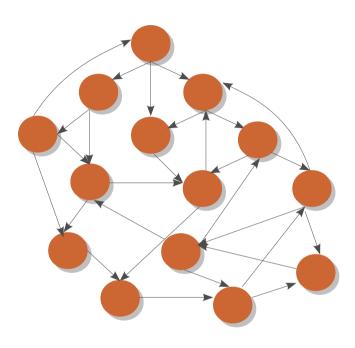


- Bring logically close graph nodes also physically close in memory
- Improves spatial locality

Outline

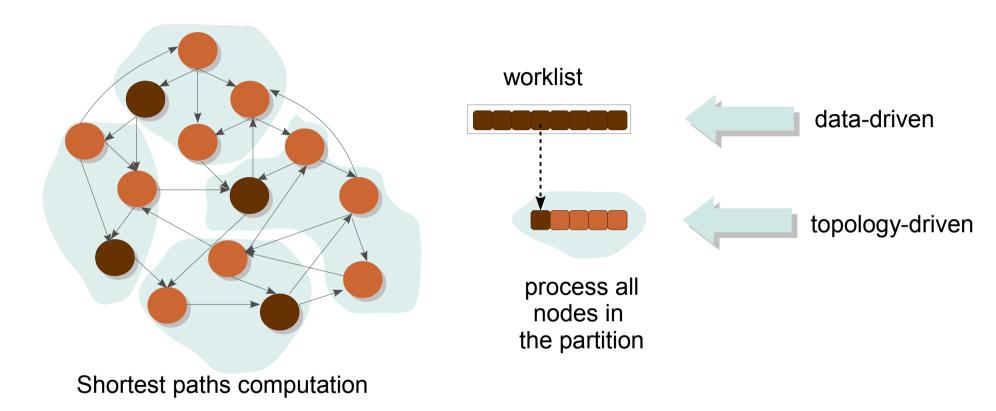
- Motivation
- Data-driven approach
- Topology-driven approach
- → Hybrid approach
 - spatial
 - temporal
- Evaluation
- Conclusions

Spatial Hybrid

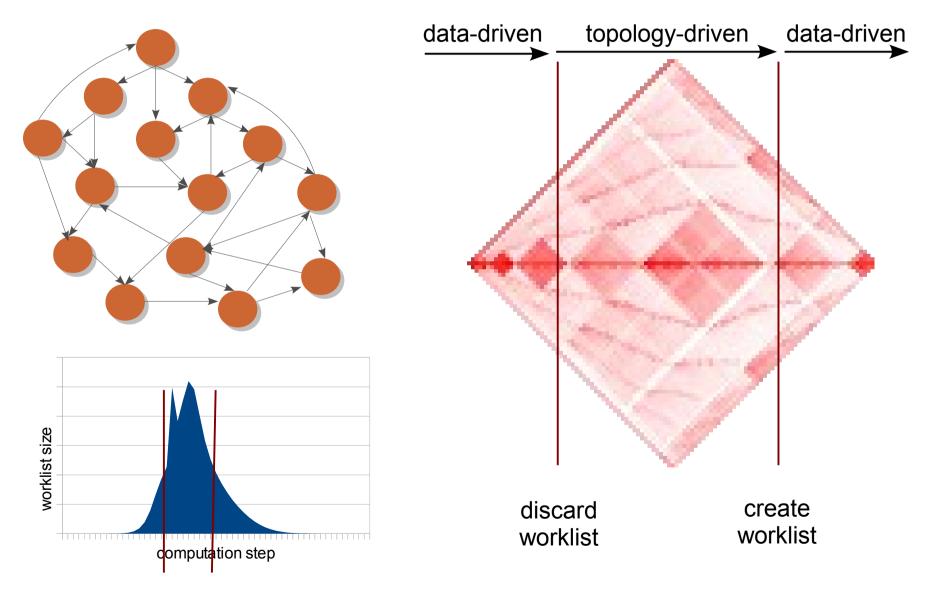


Shortest paths computation

Spatial Hybrid



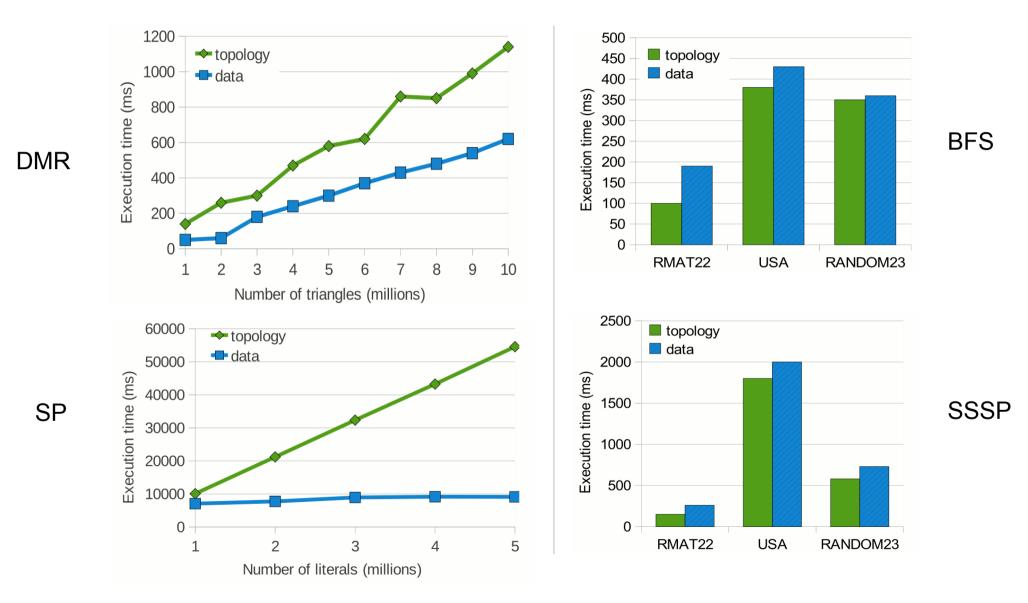
Temporal Hybrid



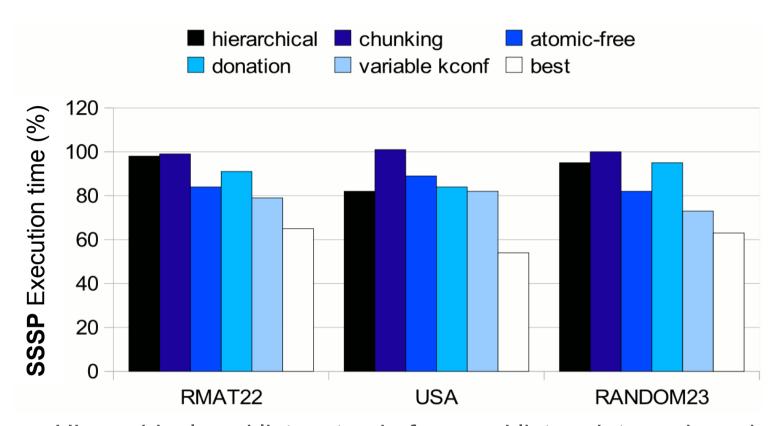
Outline

- Motivation
- Data-driven approach
- Topology-driven approach
- Hybrid approach
- → Evaluation
 - Fermi (Quadro 6000, 1.45 GHz, 448 cores over 14 SMs)
 - LonestarGPU benchmark suite
 http://iss.ices.utexas.edu/?p=projects/galois/lonestargpu
- Conclusions

Data-driven vs. Topology-driven

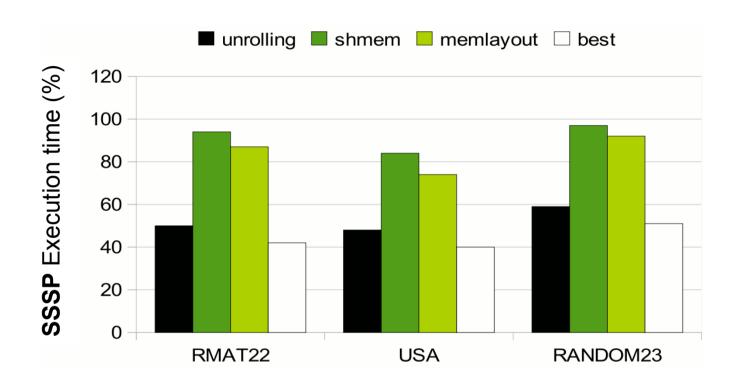


Data-driven Optimizations



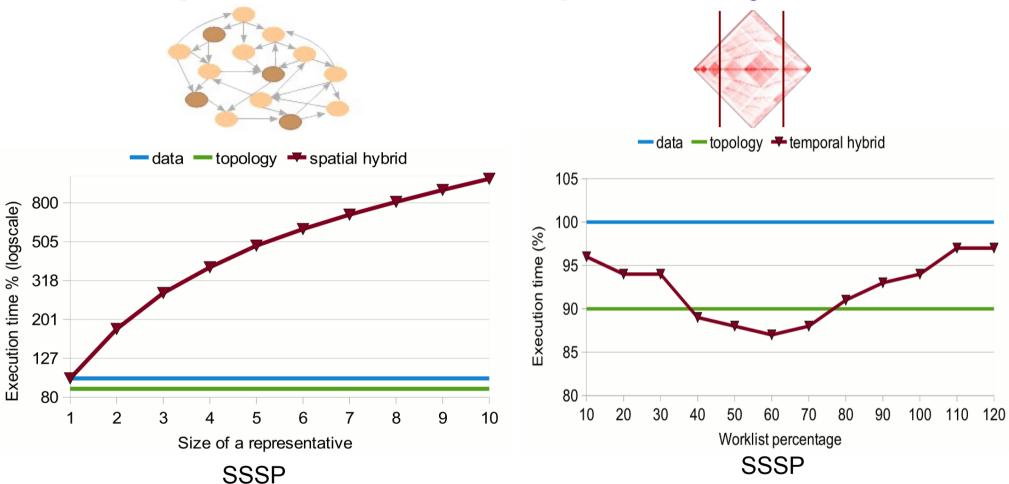
- Hierarchical worklists, atomic-free worklist update and varying configuration help
- Chunking is not effective
- Overall, performance improves by 40-50%

Topology-driven Optimizations



- Effect is more pronounced than the data-driven approach
- Unrolling helps the most
- Overall, performance improves by 50-60%

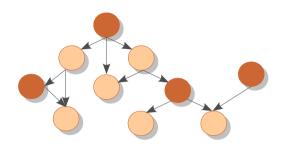
Spatial and Temporal Hybrid

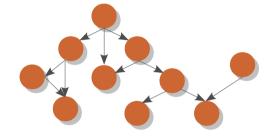


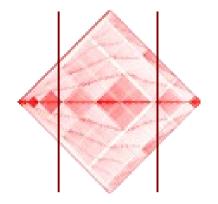
- Spatial hybrid has consistently reduced performance
- Temporal hybrid achieves performance better than the solo approaches

Conclusions

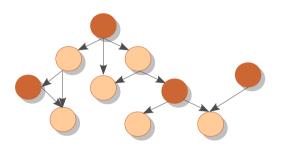
- Data-driven irregular computations usually perform better
- Topology-driven irregular computations perform better if additional algorithmic properties can be exploited
- Hybrid approaches may provide better performance than the individual approaches

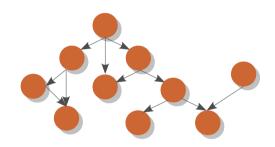






Data-driven versus Topology-driven Irregular Computations on GPUs





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LonestarGPU





DMR Performance

SP Performance

BFS Performance

SSSP Performance