

Shoal: smart allocation and replication of memory for parallel programs



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Suboptimal allocation \rightarrow bad performance

Shoal



- Memory abstraction: Arrays
- Statically derive access patterns from code
- Choose array implementation at runtime
- Reduces runtime:

- 4x over naïve memory allocation

Example: PageRank





Example: PageRank





8x8 AMD Opteron 6378 Bulldozer 4 Sockets 512 GB RAM

SNAP Twitter graph 41M nodes 1468M edges size in RAM: 2.5 GB





void *data = malloc(SIZE); memset(data, 0, SIZE);

- Implicit Linux policy on where to allocate memory
- First touch \rightarrow all memory on same NUMA node

What we would like to do?



- Partitioning
 - Split working space, put on different nodes
- Replication
 - Copy array
 - Updates: consistency
- → Reduce load-imbalance
- \rightarrow Localizes access \rightarrow reduces interconnect traffic
- DMA
- 2M/1G pages

What we have today:



- Explicit placement of memory
 - libnuma
- Advise Kernel about use of memory region

 madvise

SHOAL

Exploiting DSLs



- High-level API
- Efficient parallelization
- widely used
 - Machine learning
 - Signal processing
 - Graph processing

Idea: derive access patterns

Green Marl: graph storage







Overview: Green Marl







Modifications to Green Marl



1) Array abstraction





Array abstraction



- get() and set()
- copy_from(arr) and init_with(const)
- array_malloc(size, access_patterns)

Shoal's access patterns



- Read-only
- Sequential
- Random
- Indexed

indexed: for (i=0; i<SIZE; i++) { foo(arr[i]); } → sequential + local

2) Compiler





Derivation of access patterns



```
Foreach (t: G.Nodes) {
   Double val = k * Sum(nb: t.InNbrs){
      nb.rank / nb.OutDegree()} ;
   diff += | val - t.pg_rank |;
   t.pg_rank <= val @ t;
}</pre>
```

Derivation of access patterns



Foreach (t: G.Nodes) { Double val = k * Sum(nb: t.InNbrs){ nb.rank / nb.OutDegree()} ; diff += | val - t.pg_rank |; t.pg_rank <= val @ t;</pre>

Green Marl: graph storage







Green Marl: graph storage







Sum(nb: t.InNbrs) { // ... }; Operation: InNbrs - neighbors of node *t*:







0.2

0.1

0.2



Deriving access patterns







Deriving access patterns Systems@ETH zürich Sum(nb: t.InNbrs) { nb.rank / nb.OutDegree() 0 }; Operation: OutDegree() - of neighbor w: 1 2 nodes[nb+1] - nodes[nb] random lacksquareread-only ightarrowranks 0.2 0.1 0.2 nodes 3 0 1 r_nodes 3 0 1 edges 2 2 0 2 2 0 r edges 0 1

3) Runtime





Systems@ETH zürich start replicated y fits all nodes? indexed? read-only? n y n y n distributed partitioned H/W characteristics

Runtime: Choice of arrays



Runtime: Choice of arrays





Shoal workflow





Alternative approaches



- Search-based auto-tuning
- HW page migration
- Carrefour: Online analysis of access patterns
 - Simon Fraser University, ASPLOS 2013
 - Performance counters to monitor accesses
 - Dynamically migrate and replicate pages

EVALUATION

Single node allocation





single-node

Distribute memory





Carrefour: reactive tuning





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Shoal





- Knowledge of access patterns
- Replication
- Distribution
- Large pages (2M)

single-node distributed Carrefour Shoal

cores

Performance breakdown





Conclusion



- Memory abstraction, arrays
- Compiler analysis \rightarrow derive access patterns
- Runtime library \rightarrow selects implementation
- Works well with domain specific languages
- Also: support for manual annotation
 Too complex, too dynamic → Online
- Public release next week