

# Adventures in Load Balancing at Scale: Successes, Fizzles, and Next Steps

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#### **Outline**

- Introduction
  - Two abstract programming models
  - Load balancing and master/slave algorithms
  - A collaboration on modeling small nuclei
- The Asynchronous, Dynamic, Load-Balancing Library (ADLB)
  - The model
  - The API
  - An implementation
- Results
  - Serious GFMC: complex Monte Carlo physics application
  - Fun Sudoku solver
  - Parallel programming for beginners: Parameter sweeps
  - Useful batcher: running independent jobs
- An interesting alternate implementation that scales less well
- Future directions
  - for the API
  - yet another implementation

# Two Classes of Parallel Programming Models

#### Data Parallelism

- Parallelism arises from the fact that physics is largely local
- Same operations carried out on different data representing different patches of space
- Communication usually necessary between patches (local)
  - global (collective) communication sometimes also needed
- Load balancing sometimes needed

#### Task Parallelism

- Work to be done consists of largely independent tasks, perhaps not all of the same type
- Little or no communication between tasks
- Traditionally needs a separate "master" task for scheduling
- Load balancing fundamental



#### **Load Balancing**

- Definition: the assignment (scheduling) of tasks (code + data) to processes so as to minimize the total idle times of processes
- Static load balancing
  - all tasks are known in advance and pre-assigned to processes
  - works well if all tasks take the same amount of time
  - requires no coordination process
- Dynamic load balancing
  - tasks are assigned to processes by coordinating process when processes become available
  - Requires communication between manager and worker processes
  - Tasks may create additional tasks
  - Tasks may be quite different from one another



## Green's Function Monte Carlo - A Complex Application

- Green's Function Monte Carlo -- the "gold standard" for ab initio calculations in nuclear physics at Argonne (Steve Pieper, PHY)
- A non-trivial master/slave algorithm, with assorted work types and priorities; multiple processes create work dynamically; large work units
- Had scaled to 2000 processors on BG/L a little over four years ago, then hit scalability wall.
- Need to get to 10's of thousands of processors at least, in order to carry out calculations on <sup>12</sup>C, an explicit goal of the UNEDF SciDAC project.
- The algorithm threatened to become even more complex, with more types and dependencies among work units, together with smaller work units
- Wanted to maintain master/slave structure of physics code
- This situation brought forth ADLB
- Achieving scalability has been a multi-step process
  - balancing processing
  - balancing memory
  - balancing communication

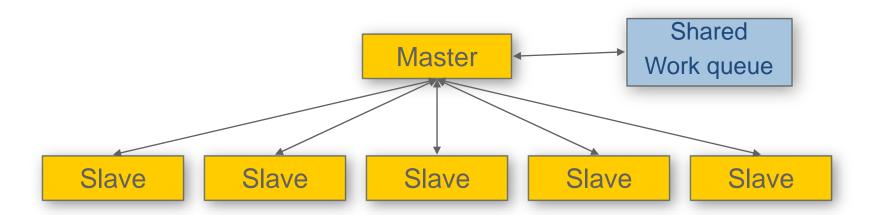


#### The Plan

- Design a library that would:
  - allow GFMC to retain its basic master/slave structure
  - eliminate visibility of MPI in the application, thus simplifying the programming model
  - scale to the largest machines



#### Generic Master/Slave Algorithm



- Easily implemented in MPI
- Solves some problems
  - implements dynamic load balancing
  - termination
  - dynamic task creation
  - can implement workflow structure of tasks
- Scalability problems
  - Master can become a communication bottleneck (granularity dependent)
  - Memory can become a bottleneck (depends on task description size)

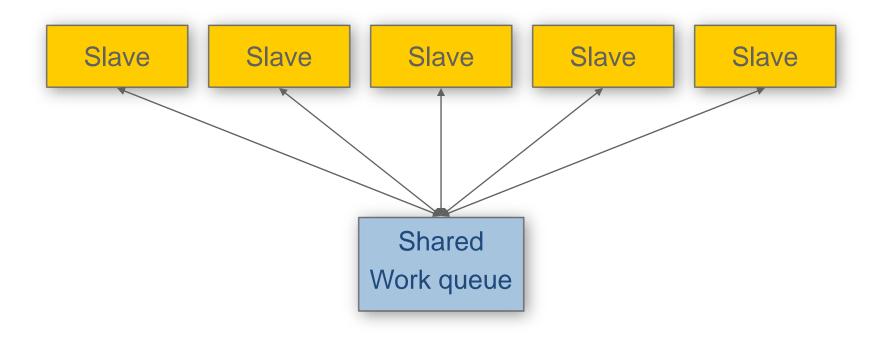


#### The ADLB Vision

- No explicit master for load balancing; slaves make calls to ADLB library; those subroutines access local and remote data structures (remote ones via MPI).
- Simple Put/Get interface from application code to distributed work queue hides MPI calls
  - Advantage: multiple applications may benefit
  - Wrinkle: variable-size work units, in Fortran, introduce some complexity in memory management
- Proactive load balancing in background
  - Advantage: application never delayed by search for work from other slaves
  - Wrinkle: scalable work-stealing algorithms not obvious



# The ADLB Model (no master)



- Doesn't really change algorithms in slaves
- Not a new idea (e.g. Linda)
- But need scalable, portable, distributed implementation of shared work queue
  - MPI complexity hidden here



# API for a Simple Programming Model

- Basic calls
  - ADLB\_Init( num\_servers, am\_server, app\_comm)
  - ADLB\_Server()
  - ADLB\_Put( type, priority, len, buf, target\_rank, answer\_dest )
  - ADLB\_Reserve( req\_types, handle, len, type, prio, answer\_dest)
  - ADLB\_Ireserve( ... )
  - ADLB\_Get\_Reserved( handle, buffer )
  - ADLB\_Set\_Done()
  - ADLB Finalize()
- A few others, for tuning and debugging
  - ADLB\_{Begin,End}\_Batch\_Put()
  - Getting performance statistics with ADLB\_Get\_info(key)

#### **API Notes**

- Return codes (defined constants)
  - ADLB\_SUCCESS
  - ADLB\_NO\_MORE\_WORK
  - ADLB\_DONE\_BY\_EXHAUSTION
  - ADLB\_NO\_CURRENT\_WORK (for ADLB\_Ireserve)
- Batch puts are for inserting work units that share a large proportion of their data
- Types, answer\_rank, target\_rank can be used to implement some common patterns
  - Sending a message
  - Decomposing a task into subtasks
  - Maybe should be built into API

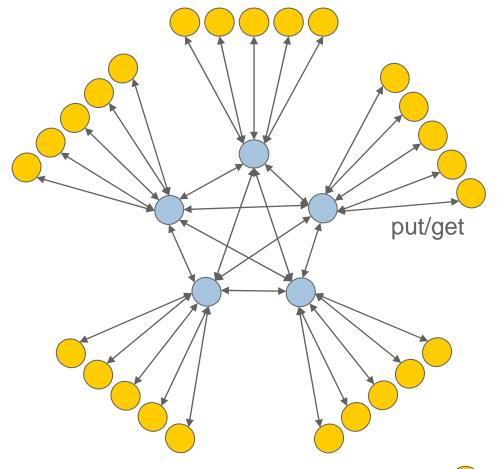


#### More API Notes

- If some parameters are allowed to default, this becomes a simple, highlevel, work-stealing API
  - examples follow
- Use of the "fancy" parameters on Puts and Reserve-Gets allows variations that allow more elaborate patterns to be constructed
- This allows ADLB to be used as a low-level execution engine for higher-level models
  - API's being considered as part of other projects



#### **How It Works**



- Application Processes
- ADLB Servers

#### Early Experiments with GFMC/ADLB on BG/P

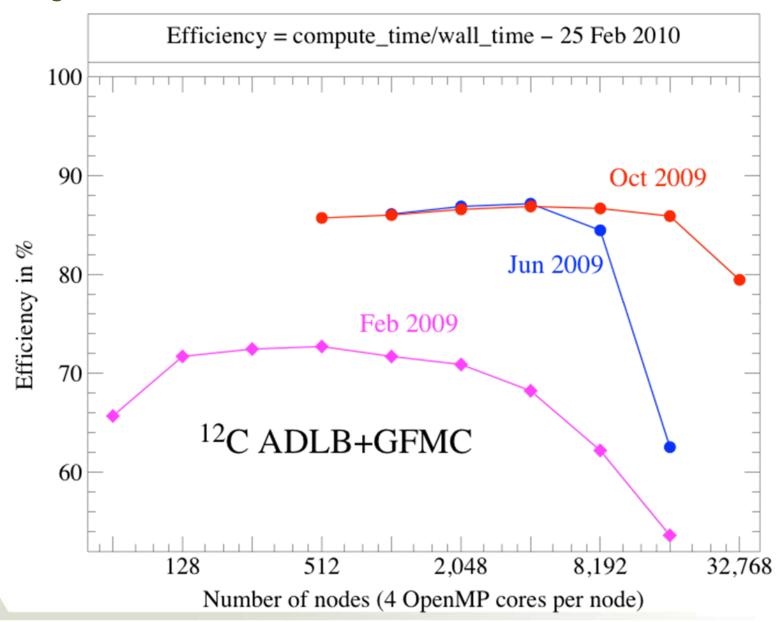
- Using GFMC to compute the binding energy of 14 neutrons in an artificial well ("neutron drop" = teeny-weeny neutron star)
- A weak scaling experiment

BG/P cores	ADLB Servers	Configs	Time (min.)	Efficiency (incl. serv.)
4K	130	20	38.1	93.8%
8K	230	40	38.2	93.7%
16K	455	80	39.6	89.8%
32K	905	160	44.2	80.4%

- Recent work: "micro-parallelization" needed for <sup>12</sup>C, OpenMP in GFMC.
  - a successful example of hybrid programming, with ADLB + MPI + OpenMP

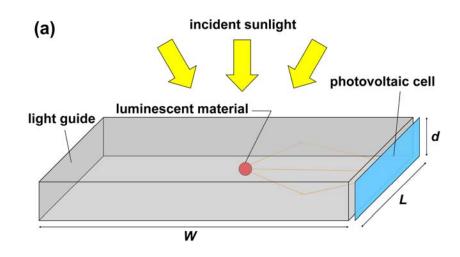


# Progress with GFMC



## **Another Physics Application - Parameter Sweep**

- Luminescent solar concentrators
  - Stationary, no moving parts
  - Operate efficiently under diffuse light conditions (northern climates)
- Inexpensive collector, concentrate light on high-performance solar cell
- In this case, the authors never learned any parallel programming approach before ADLB



#### The "Batcher"

- Simple but potentially useful
- Input is a file of Unix command lines
- ADLB worker processes execute each one with the Unix "system" call



# A Tutorial Example: Sudoku

1	2				9			7
		3				6	1	
				7		8		
					5	3		
7		9	1		8	2		6
		5	6					
		1		9				
	6	7				1		
2			5				3	8



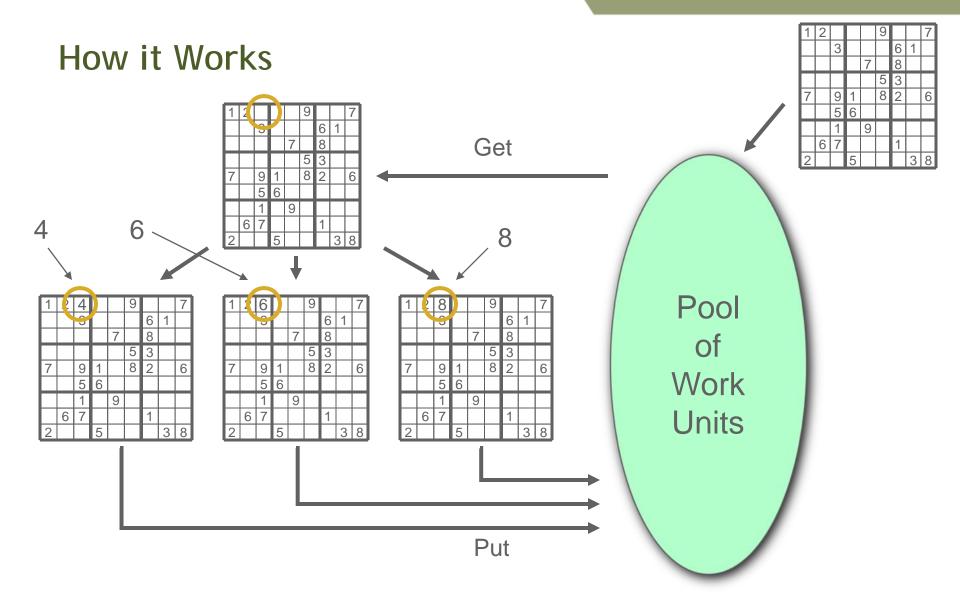
#### Parallel Sudoku Solver with ADLB

1	2				9			7
		3				6	1	
				7		8		
					5	3		
7		9	1		8	2		6
		5	6					
		1		9				
	6	7				1		
2			5				3	8

```
Work unit = partially completed "board"
```

```
Program:
 if (rank = 0)
   ADLB Put initial board
 ADLB_Get board (Reserve+Get)
 while success (else done)
  ooh
   find first blank square
   if failure (problem solved!)
       print solution
       ADLB Set Done
   else
       for each valid value
         set blank square to value
         ADLB Put new board
       ADLB Get board
end while
```





After initial Put, all processes execute same loop (no master)



## Optimizing Within the ADLB Framework

- Can embed smarter strategies in this algorithm
  - ooh = "optional optimization here", to fill in more squares
  - Even so, potentially a <u>lot</u> of work units for ADLB to manage
- Can use priorities to address this problem
  - On ADLB\_Put, set priority to the number of filled squares
  - This will guide depth-first search while ensuring that there is enough work to go around
    - How one would do it sequentially
- Exhaustion automatically detected by ADLB (e.g., proof that there is only one solution, or the case of an invalid input board)



## The ADLB Server Logic

- Main loop:
  - MPI\_Iprobe for message in busy loop
  - MPI\_Recv message
  - Process according to type
    - Update status vector of work stored on remote servers
    - Manage work queue and request queue
    - (may involve posting MPI\_Isends to isend queue)
  - MPI\_Test all requests in isend queue
  - Return to top of loop
- The status vector replaces single master or shared memory
  - Circulates every .1 second at high priority
  - Multiple ways to achieve priority



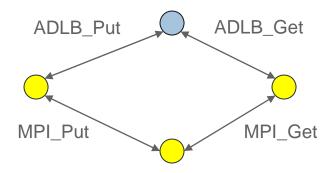
#### ADLB Uses Multiple MPI Features

- ADLB\_Init returns separate application communicator, so application can use MPI for its own purposes if it needs to.
- Servers are in MPI\_Iprobe loop for responsiveness.
- MPI\_Datatypes for some complex, structured messages (status)
- Servers use nonblocking sends and receives, maintain queue of active MPI\_Request objects.
- Queue is traversed and each request kicked with MPI\_Test each time through loop; could use MPI\_Testany. No MPI\_Wait.
- Client side uses MPI\_Ssend to implement ADLB\_Put in order to conserve memory on servers, MPI\_Send for other actions.
- Servers respond to requests with MPI\_Rsend since MPI\_Irecvs are known to be posted by clients before requests.
- MPI provides portability: laptop, Linux cluster, SiCortex, BG/P
- MPI profiling library is used to understand application/ADLB behavior.



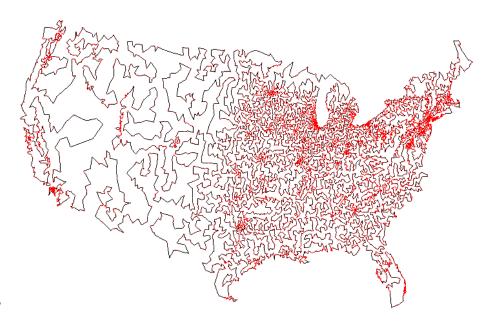
#### An Alternate Implementation of the Same API

- Motivation for 1-sided, single-server version
  - Eliminate multiple views of "shared" queue data structure and the effort required to keep them (almost) coherent)
  - Free up more processors for application calculations by eliminating most servers.
  - Use larger client memory to store work packages
- Relies on "passive target" MPI-2 remote memory operations
- Single master proved to be a scalability bottleneck at 32,000 processors (8K nodes on BG/P) not because of processing capability but because of network congestion.
- Have not yet experimented with hybrid version



# **Getting ADLB**

- Web site is <a href="http://www.cs.mtsu.edu/~rbutler/adlb">http://www.cs.mtsu.edu/~rbutler/adlb</a>
- To download adlb:
  - svn co http://svn.cs.mtsu.edu/svn/adlbm/trunk adlbm
- What you get:
  - source code (both versions)
  - configure script and Makefile
  - README, with API documentation
  - Examples
    - Sudoku
    - Batcher
      - Batcher README
    - Traveling Salesman Problem
- To run your application
  - configure, make to build ADLB library
  - Compile your application with mpicc, use Makefile as example
  - Run with mpiexec
- Problems/complaints/kudos to {lusk,rbutler}@mcs.anl.gov



#### **Future Directions**

- API design
  - Some higher-level function calls might be useful
  - User community will generate these
- Implementations
  - The one-sided version
    - implemented
    - single server to coordinate matching of requests to work units
    - stores work units on client processes
    - Uses MPI\_Put/Get (passive target) to move work
    - Hit scalability wall for GFMC at about 8000 processes
  - The thread version
    - uses separate thread on each client; no servers
    - the original plan
    - maybe for BG/Q, where there are more threads per node
    - not re-implemented (yet)



#### Conclusions

- The Philosophical Accomplishment: Scalability need not come at the expense of complexity
- The Practical Accomplishment: Multiple uses
  - As high-level library to make simple applications scalable
  - As execution engine for
    - complicated applications (like GFMC)
    - higher-level "many-task" programming models



# The End

