

Performance Evaluation of Plasma and Astrophysics Applications on Modern Parallel Vector Systems

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Abstract. The last decade has witnessed a rapid proliferation of superscalar cache-based microprocessors to build high-end computing (HEC) platforms, primarily because of their generality, scalability, and cost effectiveness. However, the growing gap between sustained and peak performance for full-scale scientific applications on such platforms has become major concern in high performance computing. The latest generation of custom-built parallel vector systems have the potential to address this concern for numerical algorithms with sufficient regularity in their computational structure. In this work, we explore two and three dimensional implementations of a plasma physics application, as well as a leading astrophysics package on some of today's most powerful supercomputing platforms. Results compare performance between the the vector-based Cray X1, Earth Simulator, and newly-released NEC SX-8, with the commodity-based superscalar platforms of the IBM Power3, Intel Itanium2, and AMD Opteron. Overall results show that the SX-8 attains unprecedented aggregate performance across our evaluated applications.

1 Introduction

The last decade has witnessed a rapid proliferation of superscalar cache-based microprocessors to build high-end computing (HEC) platforms. This is primarily because their generality, scalability, and cost effectiveness convinced computer vendors and users that vector architectures hold little promise for future large-scale supercomputing systems. However, the constant degradation of superscalar sustained performance has become a well-known problem in the scientific computing community. This trend has been widely attributed to the use of superscalar-based commodity components whose architectural designs offer a balance between memory performance, network capability, and execution rate, that is poorly matched to the requirements of large-scale numerical computations. The latest generation of custom-built parallel vector systems are addressing these challenges for numerical algorithms amenable to vectorization.

Superscalar architectures are unable to efficiently exploit the large number of floating-point units that can be potentially fabricated on a chip, due to the small

granularity of their instructions and the correspondingly complex control structure necessary to support it. Vector technology, on the other hand, provides an efficient approach for controlling a large amount of computational resources provided that sufficient regularity in the computational structure can be discovered. Vectors exploit these regularities to expedite uniform operations on independent data elements, allowing memory latencies to be masked by overlapping pipelined vector operations with memory fetches. Vector instructions specify a large number of identical operations that may execute in parallel, thus reducing control complexity and efficiently controlling a large amount of computational resources. However, when such operational parallelism cannot be found, the efficiency of the vector architecture can suffer from the properties of Amdahl’s Law, where the time taken by the portions of the code that are non-vectorizable easily dominate the execution time.

In order to quantify what modern vector capabilities entail for the scientific communities that rely on modeling and simulation, it is critical to evaluate them in the context of demanding computational algorithms. This work compares performance between the vector-based Cray X1, Earth Simulator (ES) and newly-released NEC SX-8, with commodity-based superscalar platforms: the IBM Power3, Intel Itanium2, and AMD Opteron. We study the behavior of three scientific codes with the potential to run at ultra-scale, in the areas of plasma physics (LBMHD2D and LBMHD3D), and astrophysics (CACTUS). Our work builds on our previous efforts [1, 2] and makes the contribution of adding recently acquired performance data for the SX-8, and the latest generation of superscalar processors. Additionally, we explore improved vectorization techniques for 2LBMHD and Cactus boundary conditions. Overall results show that the SX-8 attains unprecedented aggregate performance across our evaluated applications, continuing the trend set by the ES in our previous performance studies.

2 HEC Platforms and Evaluated Applications

In this section we briefly describe the computing platforms and scientific applications examined in our study. Tables 1 and 2 present an overview of the salient features for the five parallel HEC architectures. Observe that the vector machines have higher peak performance and better system balance than the superscalar platforms. Additionally, the X1, ES, and SX-8 have high memory bandwidth relative to peak CPU speed (bytes/flop), allowing them to more effectively feed the arithmetic units. Finally, the vector platforms utilize interconnects that are tightly integrated to the processing units, with high performance network buses and low communication software overhead.

Three superscalar commodity-based platforms are examined in our study. The IBM Power3 experiments reported were conducted on the 380-node IBM pSeries system, Seaborg, running AIX 5.2 (Xlf compiler 8.1.1) and located at Lawrence Berkeley National Laboratory (LBNL). Each SMP node consists of sixteen 375 MHz processors (1.5 Gflop/s peak) connected to main memory via

the Colony switch using an omega-type topology. The AMD Opteron system, Jacquard, is also located at LBNL and contains 320 dual nodes, running Linux 2.6.5 (PathScale 2.0 compiler). Each node contains two 2.2 GHz Opteron processors (4.4 Gflop/s peak), interconnected via Infiniband fabric in a fat-tree configuration. Finally, the Intel Itanium experiments were performed on the Thunder system, consisting of 1024 nodes, each containing four 1.4 GHz Itanium2 processors (5.6 Gflop/s peak) and running Linux Chaos 2.0 (Fortran version ifort 8.1). The system is interconnected using Quadrics Elan4 in a fat-tree configuration, and is located at Lawrence Livermore National Laboratory.

We also examine three state-of-the-art parallel vector systems. The Cray X1 is designed to combine traditional vector strengths with the generality and scalability features of modern superscalar cache-based parallel systems. The computational core, called the single-streaming processor (SSP), contains two 32-stage vector pipes running at 800 MHz. Each SSP contains 32 vector registers holding 64 double-precision words, and operates at 3.2 Gflop/s peak for 64-bit data. The SSP also contains a two-way out-of-order superscalar processor running at 400 MHz with two 16KB caches (instruction and data). Four SSP can be combined into a logical computational unit called the multi-streaming processor (MSP) with a peak of 12.8 Gflop/s. The four SSPs share a 2-way set associative 2MB data Ecache, a unique feature for vector architectures that allows extremely high bandwidth (25–51 GB/s) for computations with temporal data locality. The X1 node consists of four MSPs sharing a flat memory, and large system configuration are networked through a modified 2D torus interconnect. All reported X1 experiments were performed on the 512-MSP system (several reserved for system services) running UNICOS/mp 2.5.33 (5.3 programming environment) and operated by Oak Ridge National Laboratory.

The vector processor of the ES uses a dramatically different architectural approach than conventional cache-based systems. Vectorization exploits regularities in the computational structure of scientific applications to expedite uniform operations on independent data sets. The 500 MHz ES processor is an enhanced NEC SX6, containing an 8-way replicated vector pipe with a peak performance of 8.0 Gflop/s per CPU. The Earth Simulator is the world’s third most powerful supercomputer [3], containing 640 ES nodes connected through a custom

Table 1. CPU overview of the Power3, Itanium2, Opteron, X1, ES, and SX-8 platforms.

Platform	CPU/ Node	Clock (MHz)	Peak (GF/s)	Mem BW (GB/s)	Peak (Byte/Flop)
Power3	16	375	1.5	0.7	0.47
Itanium2	4	1400	5.6	6.4	1.1
Opteron	2	2200	4.4	6.4	1.5
X1	4	800	12.8	34.1	2.7
ES (Modified SX-6)	8	500	8.0	32.0	4.0
SX-8	8	2000	16.0	64.0	4.0

single-stage IN crossbar. The 5120-processor ES runs Super-UX, a 64-bit Unix operating system based on System V-R3 with BSD4.2 communication features. As remote ES access is not available, the reported experiments were performed during the authors' visit to the Earth Simulator Center located in Kanazawa-ku, Yokohama, Japan in 2003 and 2004.

Finally, we examine the newly-released NEC SX-8, the world's most powerful vector processor. The SX-8 architecture operates at 2 GHz, and contains four replicated vector pipes for a peak performance of 16 Gflop/s per processor. The SX-8 architecture has several enhancements compared with the ES/SX6 predecessor, including improved divide performance, hardware square root functionality, and in-memory caching for reducing bank conflict overheads. However, the SX-8 in our study uses commodity DDR-SDRAM; thus, we expect higher memory overhead for irregular accesses when compared with the specialized high-speed FPLRAM (Full Pipelined RAM) of the ES. Both the ES and SX-8 processors contain 72 vector registers each holding 256 doubles, and utilize scalar units operating at the half the peak of their vector counterparts. All reported SX-8 results were run on the 36 node (72 soon to be available) system located at High Performance Computer Center (HLRS) in Stuttgart, Germany. This HLRS SX-8 is interconnected with the NEC Custom IXS network and runs Super-UX (Fortran Version 2.0 Rev.313). All missing performance results will appear in the final paper version.

2.1 Scientific Applications

The application domains from scientific computing were chosen to compare the performance of the vector-based X1, ES, and SX-8 with the superscalar-based Power3, Itanium2, and Opteron systems. We examine LBMHD2D and LBMHD3D, two- and three-dimensional implementations of a plasma physics applications that use the Lattice-Boltzmann method to study magneto-hydrodynamics; and CACTUS, a modular framework supporting a wide variety of multi-physics applications [4], using the Arnowitt-Deser-Misner (ADM) Baumgarte-Shapiro-Shibata-Nakamura (BSSN) [5] method for simulation of black holes. An overview of the applications is presented in Table 3.

Table 2. Interconnect performance of the Power3, Itanium2, Opteron, X1, ES, and SX-8 platforms.

Platform	Network	MPI Lat (μ sec)	MPI BW (GB/s/CPU)	Bisect BW (Byte/Flop)	Network Topology
Power3	Colony	16.3	0.13	0.09	Fat-tree
Itanium2	Quadrics	3.0	0.25	0.04	Fat-tree
Opteron	InfiniBand	6.0	0.59	0.11	Fat-tree
X1	Custom	7.3	6.3	0.09	2D-torus
ES (Modified SX-6)	Custom (IN)	5.6	1.5	0.19	Crossbar
SX-8	IXS	5.0	2.0	0.13	Crossbar

Table 3. Overview of scientific applications examined in our study.

Name	Lines	Discipline	Methods	Structure
LBMHD2D	1,500	Plasma Physics	Magneto-Hydrodynamics, Lattice Boltzmann	Grid/Lattice
LBMHD3D	2,500	Plasma Physics	Magneto-Hydrodynamics, Lattice Boltzmann	Grid/Lattice
CACTUS	84,000	Astrophysics	Einstein Theory of GR, ADM- BSSN, Method of Lines	Grid

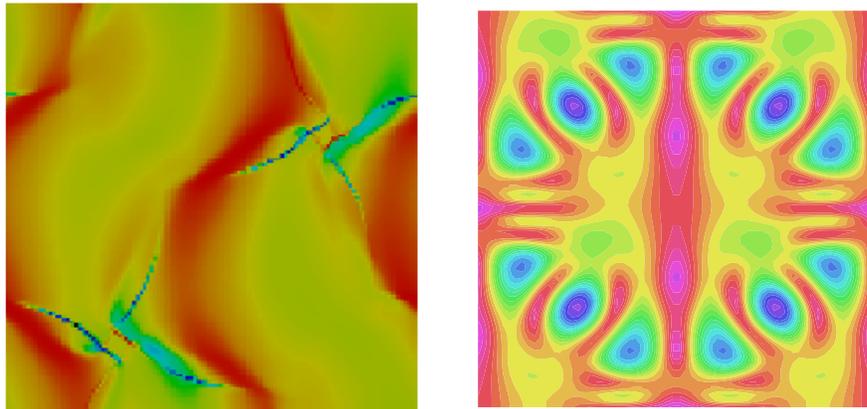
These codes represent candidate ultra-scale applications that have the potential to fully utilize leadership-class computing systems. Performance results, presented in Gflop/s per processor and percentage of peak, are used to compare the relative time to solution of our evaluated computing systems. When different algorithmic approaches are used for the vector and scalar implementations, this value is computed by dividing a valid baseline flop-count by the measured wall-clock time of each platform. Missing results will be presented in the final paper.

3 Magneto-Hydrodynamic Turbulence Simulation

Lattice Boltzmann methods (LBM) have proved a good alternative to conventional numerical approaches for simulating fluid flows and modeling physics in fluids [6]. The basic idea of the LBM is to develop a simplified kinetic model that incorporates the essential physics, and reproduces correct macroscopic averaged properties. Recently, several groups have applied the LBM to the problem of magneto-hydrodynamics (MHD) [7, 8] with promising results. We use two LB MHD codes, an previously used 2D code [9, 1] and a more recently developed 3D code. In both cases, the codes simulate the behavior of a conducting fluid evolving from simple initial conditions through the onset of turbulence. Figure 1 shows a slice through the xy-plane in the (left) 2D and right (3D) simulation, where the vorticity profile has considerably distorted after several hundred time steps as computed by LBMHD. In the 2D case, the square spatial grid is coupled to an octagonal streaming lattice and block distributed over a 2D processor grid. The 3D spatial grid is coupled via a 3DQ27 streaming lattice and block distributed over a 3D Cartesian processor grid. Each grid point is associated with a set of mesoscopic variables, whose values are stored in vectors proportional to the number of streaming directions — in this case 9 and 27 (8 and 26 plus the null vector).

The simulation proceeds by a sequence of collision and stream steps. A collision step involves data local only to that spatial point, allowing concurrent, dependence-free point updates; the mesoscopic variables at each point are updated through a complex algebraic expression originally derived from appropriate conservation laws. A stream step evolves the mesoscopic variables along

Fig. 1. Contour plot of xy-plane showing the evolution of vorticity from well-defined tube-like structures into turbulent structures using (left) LBMHD2D and (right) LBMHD3D.



the streaming lattice, necessitating communication between processors for grid points at the boundaries of the blocks.

Additionally, for the 2D case, an interpolation step is required between the spatial and streaming lattices since they do not match. This interpolation is folded into the stream step. For the 3D case, a key optimization described by Wellein and co-workers [10] was implemented, saving on the work required by the stream step. They noticed that the two phases of the simulation could be combined, so that either the newly calculated particle distribution function could be scattered to the correct neighbor as soon as it was calculated, or equivalently, data could be gathered from adjacent cells to calculate the updated value for the current cell. Using this strategy, only the points on cell boundaries require copying.

3.1 Vectorization details

The basic computational structure consists of two or three nested loops over spatial grid points (typically 1000s iterations) with inner loops over velocity streaming vectors and magnetic field streaming vectors (typically 10-30 iterations), performing various algebraic expressions. Although the two codes have kernels which are quite similar, our experiences in optimizing were somewhat different.

For the 2D case, in our earlier work on the ES, attempts to make the compiler vectorize the inner gridpoint loops rather than the streaming loops failed. The inner grid point loop was manually taken inside the streaming loops, which were hand unrolled twice in the case of small loop bodies. In addition, the array temporaries added were padded to reduce bank conflicts. With the hindsight of

our later 3D code experience, this strategy is clearly not optimal. Since more work can be inserted into the vectorized loop by unrolling the streaming loops completely, this will give better utilization of the multiple vector pipes. We have verified that this strategy does indeed give better performance than the original algorithm on both the ES and SX-8, and show results that illustrate this in the next section. Turning to the X1, the compiler did an excellent job, multi-streaming the outer grid point loop and vectorizing the inner grid point loop after unrolling the stream loops without any user code restructuring. For the superscalar architectures some effort was made to tune for better cache use. First, the inner gridpoint loop was blocked and inserted into the streaming loops to provide stride-one access in the innermost loops. The streaming loops were then partially unrolled.

For the 3D case, on both the ES and SX-8, the innermost loops were unrolled via compiler directives and the (now) innermost grid point loop was vectorized. This proved a very effective strategy, and was also followed on the X1. In the case of the X1, however, the compiler needed more coercing via directives to multi-stream the outer grid point loop and vectorize the inner grid point loop once the streaming loops had been unrolled. The difference in behavior is clearly related to the size of the unrolled loop body, the 3D case being a factor of approximately three more complicated. In a multi-streamed code the number of vector registers available for a vectorized loop is limited and for complex loop bodies register spilling will occur. However, in this case, the strategy pays off as shown experimental results section below. For the superscalar architectures, we utilized a data layout that has been previously shown to be optimal on cache-based machines [10], but did not explicitly tune for the cache size on any machine.

Interprocessor communication was implemented using the MPI library, by copying the non-contiguous mesoscopic variables data into temporary buffers, thereby reducing the required number of send/receive messages.

3.2 Experimental Results

Tables 4 and 5 and present the performance of both LBMHD applications across the five architectures evaluated in our study. Cases where the memory required exceeded that available as indicated with a dash. For LBMHD2D we show the performance of both vector algorithms (first strip-mined as used in the original ES experiment, and second the new unrolled inner) for the SX-8. In accordance with the discussion in the previous section, the new algorithm clearly outperforms the old.

Observe that the vector architectures clearly outperform the scalar systems by a significant factor. Across these architectures, the LB applications exhibit an average vector length (AVL) very close to the maximum and a very high vector operation ratio (VOR). In absolute terms, the SX-8 is the leader by a wide margin, achieving the highest per processor performance to date for LBMHD3D. The ES, however, sustains the highest fraction of peak across all architectures

Table 4. LBMHD2D performance in GFlop/s (per processor) across the studied architectures for a range of concurrencies and grid sizes. The original and optimized algorithms are shown for the ES and SX-8. Percentage of peak is shown in parenthesis.

P	Size	Power3	Itanium2	Opteron	X1	original	optimized	original	optimized
						ES	ES	SX-8	SX-8
16	4096^2	0.11 (7)	0.40 (7)	0.83 (19)	4.32 (34)	4.62 (58)	5.00 (63)	6.33 (40)	7.45 (47)
64	4096^2	0.14 (9)	0.42 (7)	0.81 (18)	4.35 (34)	4.29 (54)	4.36 (55)	4.75 (30)	6.28 (39)
64	8192^2	0.11 (7)	0.40 (7)	0.81 (18)	4.48 (35)	4.64 (58)	5.01 (62)	6.01 (38)	7.03 (44)
256	8192^2	0.12 (8)	0.38 (6)		2.70 (21)	4.26 (53)	4.43 (55)	4.44 (28)	5.51 (34)

Table 5. LBMHD3D performance in GFlop/s (per processor) across the studied architectures for a range of concurrencies and grid sizes. Percentage of peak is shown in parenthesis.

P	Size	Power3	Itanium2	Opteron	X1	ES	SX-8
16	256^3	0.14 (9)	0.26 (5)	0.70 (16)	5.19 (41)	5.50 (69)	7.89 (49)
64	256^3	0.15 (10)	0.35 (6)	0.68 (15)	5.24 (41)	5.25 (66)	8.10 (51)
256	512^3	0.14 (9)	0.32 (6)	0.60 (14)	5.26 (41)	5.45 (68)	9.66 (60)
512	512^3	0.14 (9)	0.35 (6)	0.59 (13)	—	5.21 (65)	—

— 65% even at the highest 512-processor concurrency. Examining the X1 behavior, we see that in MSP mode absolute performance is similar to the ES. The high performance of the X1 is gratifying since we noted several outputted warnings concerning vector register spilling during the optimization of the collision routine. Because the X1 has fewer vector registers than the ES/SX-8 (32 vs 72), vectorizing these complex loops will exhaust the hardware limits and force spilling to memory. That we see no performance penalty is probably due to the spilled registers being effectively cached.

Turning to the superscalar architectures, the Opteron cluster outperforms the Itanium2 system by almost a factor of 2X. One source of this disparity is that the Opteron achieves stream memory bandwidth [11] of more than twice that of the Itanium2. Another possible source of this degradation are the relatively high cost of inner-loop register spills on the Itanium2, since the floating point values cannot be stored in the first level of cache. Given the age and specifications, the Power3 does quite reasonably, obtaining a higher percent of peak than the Itanium2, but falling behind the Opteron.

Although the SX-8 achieves the highest absolute performance, the percentage of peak is somewhat lower than that of ES. We believe that this is related to the memory subsystem and use of DDR-SDRAM. In order to try to test this hypothesis, we recorded the time due to memory bank conflicts for both applications on the ES and SX-8 using the ftrace tool, and present it in Table 6.

Table 6. LBMHD2D and LBMHD3D bank conflict time (as percentage of real time) shown for a range of concurrencies and grid sizes on ES and SX-8.

<i>Code</i>	<i>P</i>	Grid Size	ES BC (%)	SX-8 BC (%)
2D	64	8192^2	0.3	16.6
2D	256	8192^2	0.3	10.7
3D	64	256^3	>0.01	1.1
3D	256	512^3	>0.01	1.2

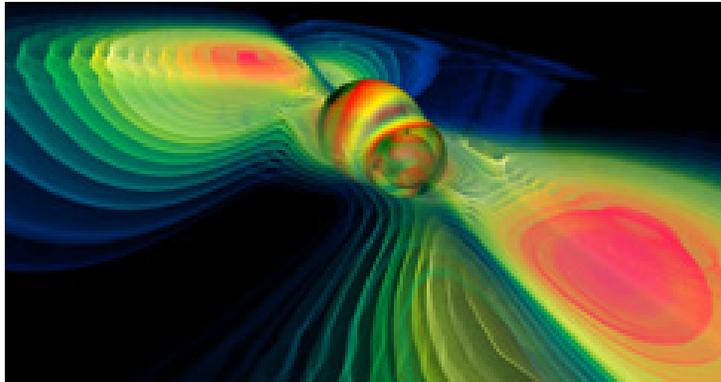
Most obviously in the case of the 2D code, the amount of time spent due to bank conflicts is appreciably larger for the SX-8. Efforts to reduce the amount of time for bank conflicts for the 2D 64 processor benchmark produced a slight improvement to 13%. In the case of the 3D code, the effects of bank conflicts are very minimal.

4 CACTUS

One of the most challenging problems in astrophysics is the numerical solution of Einstein’s equations following from the Theory of General Relativity (GR): a set of coupled nonlinear hyperbolic and elliptic equations containing thousands of terms when fully expanded. The Cactus Computational ToolKit [12, 5] is designed to evolve Einstein’s equations stably in 3D on supercomputers to simulate astrophysical phenomena with high gravitational fluxes – such as the collision of two black holes and the gravitational waves radiating from that event. While Cactus is a modular framework supporting a wide variety of multi-physics applications [4], this study focuses exclusively on the GR solver, which implements the Arnowitt-Deser-Misner (ADM) Baumgarte-Shapiro-Shibata-Nakamura (BSSN) [5] method for stable evolutions of black holes. Figure 2 presents a visualization of one of the first simulations of the grazing collision of two black holes computed by the Cactus code. The merging black holes are enveloped by their “apparent horizon”, which is colorized by its Gaussian curvature. The concentric surfaces that surround the black holes are equipotential surfaces of the gravitational flux of the outgoing gravity wave generated by the collision.

The Cactus General Relativity components solve Einstein’s equations as an initial value problem that evolves partial differential equations on a regular grid using the method of finite differences. The core of the General Relativity solver uses the ADM formalism, also known also as the 3+1 form. For the purpose of solving Einstein’s equations, the ADM solver decomposes the solution into 3D spatial hypersurfaces that represent different slices of space along the time dimension. In this formalism, the equations are written as four constraint equations and 12 evolution equations. Additional stability is provided by the BSSN modifications to the standard ADM method [5]. The evolution equations can be solved using a number of different numerical approaches, including staggered

Fig. 2. Visualization of grazing collision of two black holes as computed by Cactus¹.



leapfrog, McCormack, Lax-Wendroff, and iterative Crank-Nicholson schemes. A “lapse” function describes the time slicing between hypersurfaces for each step in the evolution. A “shift metric” is used to move the coordinate system at each step to avoid being drawn into a singularity. The four constraint equations are used to select different lapse functions and the related shift vectors. For parallel computation, the grid is block domain decomposed so that each processor has a section of the global grid. The standard MPI driver for Cactus solves the PDE on a local grid section and then updates the values at the ghost zones by exchanging data on the faces of its topological neighbors in the domain decomposition.

4.1 Vectorization Details

For the superscalar systems, the computations on the 3D grid are blocked in order to improve cache locality. Blocking is accomplished through the use of temporary ‘slice buffers’, which improve cache reuse while modestly increasing the computational overhead. On vector architectures these blocking optimizations were disabled, since they reduced the vector length and inhibited performance. The ES compiler misidentified some of the temporary variables in the most compute-intensive loop of the ADM-BSSN algorithm as having inter-loop dependencies. When attempts to force the loop to vectorize failed, a temporary array was created to break the phantom dependency.

Another performance bottleneck that arose on the vector systems was the cost of calculating radiation boundary conditions. The cost of boundary condition enforcement is inconsequential on the microprocessor based systems, however they unexpectedly accounted for up to 20% of the ES runtime and over 30% of the X1 overhead. The boundary conditions were vectorized using very

¹ Visualization by Werner Bengert (AEI/ZIB) using Amira [13]

lightweight modifications such as inline expansion of subroutine calls and replication of loops to hoist conditional statements outside of the loop. Although the boundaries were vectorized via these transformations, the effective AVL remained infinitesimally small. Obtaining longer vector lengths would have required more drastic modifications that were deemed impractical due the amount of the Cactus code that would be affected by the changes. This modification was very effective on the X1 because the loops could be multistreamed. Multistreaming enabled an easy 3x performance improvement in the boundary calculations that reduced their runtime contribution from the most expensive part of the calculation to just under 9% of the overall wallclock time. These same modifications produced no net benefit for the ES or SX-8, however, because the extremely short vector lengths.

4.2 Experimental Results

The full-fledged production version of the Cactus ADM-BSSN application was run on each of the architectures with results for two grid sizes shown in Table 7. The problem size was scaled with the number of processors to keep the computational load the same (weak scaling). Cactus problems are typically scaled in this manner because their science requires the highest-possible resolutions.

For the vector systems, Cactus achieves almost perfect VOR (over 99%) while the AVL is dependent on the x-dimension size of the local computational domain. Consequently, the larger problem size (250x64x64) executed with far higher efficiency on both vector machines than the smaller test case (AVL = 248 vs. 92), achieving 34% of peak on the ES. The oddly shaped domains for the larger test case were required because the ES does not have enough memory per node to support a 250^3 domain. This rectangular grid configuration had no adverse effect on scaling efficiency despite the worse surface-to-volume ratio. Additional performance gains could be realized if the compiler was able to fuse the X and Y loop nests to form larger effective vector lengths. Also, note that for the Cactus simulations, bank conflict overheads are negligible for the chosen (not power of two) grid sizes.

Recall that the boundary condition enforcement was not vectorized on the ES and accounts for up to 20% of the execution time, compared with less than

Table 7. Cactus performance in GFlop/s (per processor) on 80x80x80 and 250x64x64 grids shown for a range of concurrencies. Percentage of peak is shown in parenthesis.

P	Size	Power3	Itanium2	Opteron	X1	ES	SX-8
16	80^3	0.31 (21)	0.60 (11)	0.96 (22)	0.54 (4)	1.47 (18)	1.86 (12)
64	80^3	0.22 (14)	0.58 (10)		0.43 (3)	1.36 (17)	
256	80^3	0.22 (14)	0.58 (10)		0.41 (3)	1.35 (17)	1.75 (11)
16	250×64^2	0.10 (6)	0.58 (10)	0.96 (22)	0.81 (6)	2.83 (35)	4.27 (27)
64	250×64^2	0.08 (6)	0.57 (10)		0.72 (6)	2.70 (34)	
256	250×64^2	0.07 (5)	0.55 (10)		0.68 (5)	2.70 (34)	3.87 (24)

5% on the superscalar systems. This demonstrates a different dimension of architectural balance that is specific to vector architectures: seemingly minor code portions that fail to vectorize can quickly dominate the overall execution time. The architectural imbalance between vector and scalar performance was particularly acute of the X1, which suffered a much greater impact from unvectorized code than the ES. On the SX-8, the boundary conditions occupy approximately the same percentage of the execution time as it did on the ES, which is consistent with the fact that the performance improvements in the SX8 scalar execution unit have scaled proportionally with the vector performance improvements. The decreased execution efficiency is primarily reflected in lower efficiency in the vector execution.

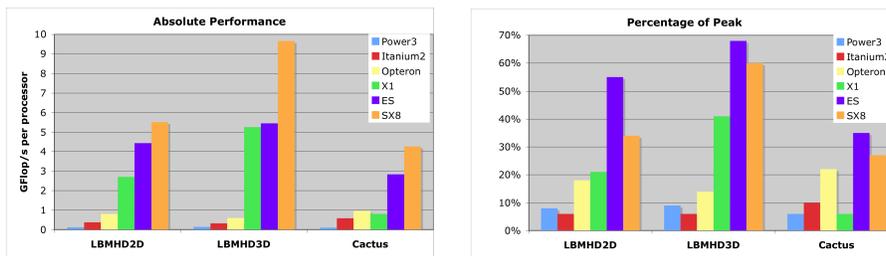
The microprocessor based systems offered lower peak performance and generally lower efficiency than the NEC vector systems. The Opteron, however, offered impressive efficiency as well as peak performance in comparison to the Power3 and the Itanium2. Unlike the Power3, the Opteron maintains its performance even for the larger problem size. The relatively low scalar performance on the microprocessor-based systems is partially due to register spilling, which is caused by the large number of variables in the main loop of the BSSN calculation. However, the much lower memory latency of the Opteron and higher effective memory bandwidth relative to its peak performance allow it to maintain higher efficiency than the Itanium2 or the Power3 processor.

In terms of communication overhead, the ES and Itanium2 systems spend less than 13% of the overall wallclock time in MPI communication compared with 23% on the Power3; highlighting the superior architectural balance of the network design of the crossbar used for the ES and Quadrics QSNNet which is used on the Itanium2 system. The final paper version will contain the complete set of Opteron/InfiniBand scaling results and corresponding analysis.

5 Conclusions

This study examined three scientific codes on the parallel vector architectures of the X1, ES and SX-8, and three superscalar platforms, Power3, Itanium2, and

Fig. 3. Summary comparison of (left) raw performance and (right) percentage of peak across our set of evaluated applications and architectures.



Opteron. A summary of the results for the largest comparable problem size and concurrency is shown in Figure 3, for both (left) raw performance and (right) percentages of peak. Overall results show that the SX-8 achieved the highest performance of any architecture tested to date, demonstrating the tremendous potential of modern parallel vector systems. However, the SX-8 could not match the sustained performance of the ES, due in part, to a relatively higher memory latency overhead for irregular data accesses. Both the SX-8 and ES also consistently achieved a significantly higher fraction of peak than the X1, due to superior scalar processor performance, memory bandwidth, and network bisection bandwidth relative to the peak vector flop rate. Finally, a comparison of the superscalar platforms shows that the Opteron consistently outperformed the Itanium2 and Power3, both in terms of raw speed and efficiency - due, in part, to its on-chip memory controller and (unlike the Itanium2) the ability to store floating point data in the L1 cache. The Itanium2 exceeds the performance of the (relatively old) Power3 processor, however its obtained percentage of peak falls further behind. Future work will expand our study to include additional areas of computational sciences, while examining the latest generation of supercomputing platforms, including BG/L, X1E, and XT3.

Acknowledgments

The authors would like to thank the staff of the Earth Simulator Center, especially Dr. T. Sato, S. Kitawaki and Y. Tsuda, for their assistance during our visit. We are also grateful for the early SX-8 system access provided by HLRS, Germany. This research used resources of the National Energy Research Scientific Computing Center, which is supported by the Office of Science of the U.S. Department of Energy under Contract No. DE-AC03-76SF00098. This research used resources of the Lawrence Livermore National Laboratory, which is supported by the Office of Science of the U.S. Department of Energy under contract No. W-7405-Eng-48. This research used resources of the Center for Computational Sciences at Oak Ridge National Laboratory, which is supported by the Office of Science of the Department of Energy under Contract DE-AC05-00OR22725. LBNL authors were supported by the Office of Advanced Scientific Computing Research in the Department of Energy Office of Science under contract number DE-AC03-76SF00098.

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