Performance of a Lattice Quantum Chromodynamics Kernel on the Cell Processor

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Abstract

The implementation of a proof-of-concept Lattice Quantum Chromodynamics kernel on the Cell processor is described in detail, illustrating issues encountered in the porting process. The resulting code performs up to 45GFlop/s per socket, indicating that the Cell processor is likely to be a good platform for future Lattice QCD calculations.

Key words: Cell processor, Multi-core programming, Lattice QCD
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1. Introduction

The Cell processor is a heterogeneous multi-core processor[1], originally designed for high-bandwidth media applications, but also having potential applications in traditional HPC fields such as computational science. Interest in computational science on the Cell processor[2,3,4] is driven by its high peak floating point capability of up to 256GFlop/s in single precision. Cell-based systems are available in a blade server form factor, making possible massively parallel machines such as the RoadRunner petaflop supercomputer currently under construction[5].

In order to exploit the power of the Cell hardware, it is necessary to overcome challenges associated with the programming model, particularly the ways in which it differs from conventional processors. The greatest difference is the Cell memory model, which replaces caches with software-managed local stores and DMA engines. This leads to a requirement for explicit communication between processing cores and main memory not found in conventional multi-core programs. Another significant obstacle to writing code for the Cell or porting code to it is the SIMD instruction set used, which must be taken into account in order to obtain good floating point performance. To some extent this difficulty may be mitigated by development tools, but as explained in section 2.3, hand-optimised Cell-specific code is needed at present.

Since the details of the Cell architecture have been well explained in other publications[1,6], here we provide only a brief summary. The Cell processor consists of nine cores: one Power Processing Element (PPE) and eight Synergistic Processing Elements (SPE). The PPE is a conventional PowerPC processor, including a 512kB L2 cache. The SPEs are SIMD processors, optimised to operate on 128 bit vectors. Each SPE has a 256kB local store (LS) and associated DMA engine with access to main memory as well as the LS of other SPEs. The fabric connecting processing elements is a bus with a peak bandwidth of 204.8GB/s, with 25.6GB/s ports for each processing element and for the memory controller. The high bandwidth of the bus leaves the bandwidth to the main memory controller as the bottleneck in data movement.

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The vast majority of floating point power lies in the SPEs. In the best case, each of the 8 SPEs is capable of performing one 128 bit fused multiply-add operation per cycle, corresponding to eight single precision floating point operations, rendering a peak performance of 25.6GFlop/s per SPE on a 3.2GHz processor. Any code which is to perform well on the Cell processor must scale well enough to use all of the SPEs, and be written with a view to SIMD operations. Although not all applications can satisfy these criteria, those based on linear algebra operations tend to map well to parallel systems and to SIMD instruction sets.

To illustrate the issues associated with obtaining high performance on the Cell processor, this paper describes the process of porting a computational kernel from Lattice Quantum Chromodynamics (Lattice QCD). Lattice QCD simulations are extremely computationally demanding, to such a degree that massively parallel machines have been custom-built solely to run such calculations\[^7,8\]. This paper is limited in scope to programs on a single Cell machine, but demonstrating the per-Cell performance possible for Lattice QCD is an important step toward further work on a multi-node Cell system.

2. Implementation

Prior to the work described here, various methods for Cell programming were experimented with by the authors\[^9\]. The implementation presented here ignores issues of ease-of-programming and maintainability, in favor of seeking the maximum possible performance.

The SPE management library libspe2\[^10\], included in the Cell SDK, provides basic threading functionality loading code onto an SPE and executing it. All synchronisation and data movement is left up to the programmer. This low level approach provides great flexibility, but also demands relatively complex code to manage the SPEs, as well as complex code in the SPE programs to perform DMA operations and synchronisation as well as the actual task underway.

Creating a SPE thread using libspe2 requires only a few straightforward function calls. Firstly, a context is created (spe_context_create), where a context represents one physical SPE. Next, a SPE binary is loaded (spe_program_load) to the context, and finally the SPE program is run (spe_context_run). The SPE program is compiled with a separate toolchain to the PPE program, and then embedded in the PPE executable.

Although it is possible to implement Cell-accelerated versions of each stage in the linear algebra procedure, and call them sequentially from existing procedures, this approach cannot yield good performance, since the bandwidth to main memory is vastly outstripped by the combined floating point power of the SPEs. To make the most of the floating point power of the SPEs, it is necessary to persistently distribute the dataset across the SPEs.

2.1. Wilson-Dslash

Lattice QCD is a discretised form of a field theory, calculated on a 4D space-time lattice. Here, we refer to the spatial size of the lattice as \(L\) and the size in time as \(T\), such that the lattice has \(L^3T\) sites in total.

A common benchmark which forms the most computationally intensive part of lattice QCD calculations is the the inversion of the Dirac operator, a procedure known as Dslash \[^11,12\]. In this paper, the physical or mathematical meaning of these operations is not discussed: all that is done is to directly transform an existing scalar Dslash procedure into a Cell-optimised version. A special-case Dslash routine is presented which operates only in single precision, only on 4 dimensional systems, and only on lattice data laid out in a red-black checkerboard. We base our Cell-optimised Dslash on a reference implementation provided in the freely available QDP++ package\[^13\]. In this reference implementation, operations between neighbouring lattice sites are represented by operations between relatively shifted instances of the lattice data. To write the Cell version of the code, all these operations are unrolled into explicit calculations and DMA operations.

The Dslash procedure takes as input a spinor \(\Psi\) (a \(4 \times 3\) complex array at each lattice site), and four color matrices \(u\) (3 \(\times 3\) complex arrays at each lattice site), one for each dimension. The output is the spinor \(\chi\). Intermediate storage is needed for half-spinors (a \(2 \times 3\) complex array at each lattice site), which represent the so-called projected form of \(\Psi\).

2.2. Data arrangement

A one-dimensional block decomposition over the SPEs (of which we have \(N_{\text{SPE}}\) in total) in the time
dimension is used to divide the lattice into a $T/N_{\text{SPE}}$ thick slice stored in the LS of each SPE. This is inherently load-balanced as long as $T$ is a multiple of the number of SPEs. This is not a problematic requirement, since existing lattice QCD simulations frequently use power-of-two sizes. For convenience, let $T_{\text{chunk}} = T/N_{\text{SPE}}$.

The operands to the Dslash operation are contiguous in main memory when using the default QDP++ red-black checkerboard indexing scheme, wherein the slowest varying index is the checkerboard (0 or 1), and the next slowest varying index is the time coordinate. The calculation of an element of Chi on the SPE requires the surrounding Psi values. A series of index manipulations are performed to locate these values in the transferred buffers, which does not need to be particularly efficient, since it is done once at initialisation, and pointers to the data required for each Chi element are stored. This look-up table sacrifices local store space for computational efficiency and the development time saved by not having to optimise the index manipulations.

The operands are transferred to and from the SPE local stores using the SPE DMA engines. Issuing DMA instructions is a straightforward low-overhead operation, exposed by the SDK as C intrinsics\cite{14}. The addresses in main memory from which operands are transferred are contained in a structure which is loaded by the SPE threads from an address passed at initialisation by the PPE thread.

A beneficial characteristic of the Dslash procedure is that it is typically used repeatedly for the same value of the gauge field $U$, meaning that the $U$ data need only be transferred once. As such, the time taken to transfer $U$ is negligible when the procedure is run many times and hence the data transferred at each Dslash call is limited to the Psi input and Chi output. There is some scope for hiding communications time by double buffering, that is performing some calculation before Psi is completely loaded. This is not done here, but ad-hoc measurements of communications time vs. calculation time indicate that any gains from double buffering would not drastically alter the final performance result.

2.3. Padding

Since a color matrix (a $3 \times 3$ complex matrix) has size 72 bytes in single precision, if the first element in an array of color matrices is aligned to a 16B boundary then the subsequent element would be offset by 8 bytes, (72\%16 = 8). This causes a problem for SIMD operations, since the SPE load and store instructions only work on 16 byte boundaries. To solve this, one may either check the alignment of a color matrix prior to operating on it, and copy it out to an aligned location if necessary, or one may pad each element by 8 bytes.

Since the color matrices are loaded only once at initialisation, the cost of padding the color matrix arrays on the SPE is negligible when Dslash is called many times. Conversely, the cost of making aligned copies when operating on the color matrix is incurred at each invocation of the Dslash procedure. The padding clearly incurs a penalty in the local store space required: an increase from 72 to 80 bytes per color matrix, or 11%. The color matrices constitute the majority of the local store space used for data, so this increase is significant to the overall local store requirement. This is only a worthwhile sacrifice as long as the serial performance of operations on the SPE is a bottleneck: if the procedure were communications-limited then it might be preferable to make per-operation aligned temporaries of an unpadded array and thus accommodate a larger sublattice.

2.4. SIMD arithmetic

SIMD-aware code is necessary to obtain even reasonable performance on the SPE, because SPE load/store instructions operate on 16 byte vectors, and only at 16 byte aligned locations in the local store. To load a vector spanning a 16 byte boundary, two vector loads are required, followed by a shuffle operation to compose the desired vector in a register. This is vastly slower than loading a properly aligned vector, which is accomplished in one load instruction (6 cycles). Similarly, to perform arithmetic operations on a scalar value and extract a scalar result requires a series of shuffle operations in addition to the arithmetic instruction.

SIMD operations are exposed as intrinsic functions in the SPE C++ compiler\cite{14}, operating on 16 byte vector datatypes. These allow low-level SIMD programming without the need to write assembly language. To perform operations on permutations of the data other than the native layout, the $\text{spu\_shuffle}$ function is used. This can rearrange the data in two input vectors to arbitrary positions across two output vectors. This is used extensively in the Dslash operation, along with the usual arith-
ometric operations and the fused multiply-add provided by the SPE’s FPU).

The Dslash operation has three main stages: spin projection, multiplication by the color matrix, and spin reconstruction. Each spin projection and reconstruction function has eight variants, corresponding to forwards and backwards in each dimension. Although mathematically each of these variants is simply a matrix multiplication by a different matrix, in practice they are programmatically distinct since the matrices in question are constant, small and sparse, so the implementation of the multiplication is unrolled into the operations required for the non-zero elements. The color matrix multiplication has two variants, one for multiplying a vector by the color matrix and another for multiplying by its adjoint. Each of these 10 operations is hand-translated from scalar code in QDP++ to SIMD intrinsics for the SPE.

Optimisation of the SIMD code is facilitated by the IBM Assembly Visualizer[^1], for the Cell. This tool provides a graphical display of the execution of compiled code on a SPE, illustrating pipeline stalls and data dependencies. The ability to see the characteristics of the code without running it provides a very rapid turnaround and takes much of the guesswork out of optimisation. To obtain a more detailed view of execution, the IBM Full System Simulator[^2] may be used to model the SPE with cycle accuracy. The simulator is also useful for fine-grained benchmarking of SPE code since the simulated execution environment provides convenient access to cycle counters.

As an example of the optimisation process, the forward spin projection phase of the calculation is considered. Compiling the reference C implementation in QDP++ using `spuxlc -O3 -qhot` and using cycle counters in the Cell simulator gives a timing of 141335 cycles to project 1024 spinors. Rewriting the arithmetic as a series of `spu_madd` and `spu_shuffle` calls gives a modest runtime decrease to 98291 cycles. Using `register vector float` type temporaries to allow the compiler to reorder loads and stores more aggressively reduces the timing further to 21084 cycles, or about 21 cycles per spinor. In the code generated from the final version, all the multiply-add, shuffle, load and store instructions are pipelined.

The SPE has a large register file, with 128 registers of 16 bytes each, easily enough to accommodate a color matrix (5 registers), a spinor (6 registers) and many temporaries. However, the compiler does not always take advantage of this, particularly it does not tend to keep values in registers between subsequent inlined functions. To obtain the better performance this is overcome by fusing functions which re-use the same data. Rather than having separate projection, matrix multiplication and reconstruction functions, the three are combined, such that all three operations can share the same register temporaries. This requires 8 fused functions, one for each direction. Since the color matrix multiplication functions (plain and adjoint) are each common to 4 of the fused functions, they are implemented as pre-processor macros, operating on register vectors floats defined in the scope into which they are included. Although this is substantially less elegant and flexible than having separate functions which the compiler inlines, keeping the data in registers avoids several spurious loads and stores.

The SPE’s fused multiply-add instruction has the same cost as a single add or multiply, so when doing a multiply an extra add is effectively free, and vice-versa. This is exploited in the Dslash procedure by implementing mixtures of subtraction and addition as multiplications by arrays of ±1 combined with an addition. To retain a fair comparison to the original code, these extra multiplications by ±1 are not counted in the Flop/s. Using this scheme, our optimised spin projection function obtains a performance of 1864MFlop/s per SPE, compared to the original scalar code’s 278MFlop/s.

The process of optimising for the Cell is utterly destructive of the original, producing code whose operation is difficult to understand by inspection. As such, these Cell-optimised versions must always be maintained separately to generic code. This fits into the existing structure of QDP++, which already includes versions of key functions optimised for the SSE architecture, in addition to the generic implementation.

2.5. Extension notes

This implementation runs only on a single cell blade. If the Cell were to be used ‘in anger’ for Lattice QCD then the code would have to interleave off-node communications with on-Cell calculations and communications. The current on-Cell parallelisation where by the division between SPEs is in the T dimension could be retained in a multi-node version of the code, keeping the T dimension on-node and diminishing requirements for the inter-node intercon-
nect to a 3D torus. If the dataset was permanently distributed across SPEs rather than being scattered and gathered at each iteration, it would be natural to have the off-node communications controlled by the SPE threads, rather than involving the PPE. In fact, given that the Cell has I/O interfaces directly connected to the EIB, it is possible that the network interface would be directly connected to one of these, meaning that the data-flow could pass directly from the SPE to the network card, bypassing main memory entirely, and ending up directly on an SPE on another node. Whether this could be implemented effectively is a question for those engineering the communications hardware and software stack.

In the current code, communication in the T dimension is implicit in the distribution and collection of the fermion (Chi and Psi) data from and to main memory at each Dslash call. However, if no further manipulations to the data are required in between Dslash calls then it would be more efficient to keep this data on the SPEs. In this case, it would be necessary to implement halo swapping in the T direction in between the SPEs. Once this is the case, the PPE would be largely uninvolved apart from setting the SPEs running for a given number of iterations. In this case, it may be possible to save some amount of synchronisation runtime by implementing a peer-to-peer synchronisation between SPEs while iterating the Dslash procedure. A higher performance barrier may also be motivated by more frequent synchronisation, if off-node communications demanded multiple synchronisation points during each Dslash procedure.

Transfers larger than 16kB are currently performed by enqueuing multiple DMA transfers of up to 16kB each. Alternatively, one could compose a DMA list of such transfers and reduce overheads by generating this list at startup and enqueuing it at each iteration of the algorithm. This may yield superior performance.

The performance of the Cell implementation is such that a Dslash over a (6,4) sub-lattice takes 33µs per site, or 28.5µs in total. Using existing benchmarks of off-the-shelf interconnects\[15\], one can compare this on-chip runtime to inter-node communications in a multi-node system. Current off-the-shelf interconnects offer latencies of the order a few µs, while the size of the halo data for a sub-lattice this size is of the order tens of kB, which may be expected to incur a communications time around 10-20µs in the best case. This communications time is of the order of the serial communications time, so for best sustained performance in a multi-node system it would be imperative to overlap off-node computation and calculation.

Currently there is a fairly low ceiling on the size of sub-lattice which may be accommodated on each SPE. This limit could be made more flexible at the cost of increased communications by streaming part or all of the $U$ arrays at each iteration rather than requiring the whole dataset to fit into the LS at one time. Equally, the Psi and Chi data could be treated in a streaming manner rather than having full-sized local arrays in the LS for the whole sub-lattice assigned to the SPE. Some quantity of the added communications cost could be hidden using double-buffering. This increased flexibility would be particularly relevant in a parallel machine which may not necessarily have enough processors for each Cell to have such as small portion of the lattice as is currently required.

### 3. Performance

All benchmarks were run on a QS20 Cell blade, with two Cell processors and 1GB of memory for each processor. The two processors in the QS20 blade are configured such that their fabrics are combined to give seamless access to all SPEs from one PPE program. Benchmarks are run on a varying number of SPEs, denoted by $N_{\text{SPE}}$, from 2 to 16 (recall that each Cell processor has 8 SPEs). Each timing result is an average over 3 runs. Where the system size is varied, the spatial size is denoted $L$ and the time dimension $T$, such that there are in total $L^3 T$ sites. As a shorthand, system sizes are written $(L, T)$.

Performing typical strong scaling benchmarks is complicated by the limitation on the data size which can be accommodated in the local stores. A system size which will can be run on two SPEs is somewhat inadequate to test the performance of 16. For this reason a variety of system sizes were used, some providing coverage of low numbers of SPEs, some testing larger numbers of SPEs. In addition, weak scaling was measured using a system size with $L = 2$ and $T$ proportional to the number of SPEs used. This provides a direct test of the synchronisation and communications, since the amount of serial work per SPE remains constant.

The results of these benchmarks are shown in Fig. 1. Each series shows a different system size, including the weak scaling case in which the system size is...
proportional to the number of SPEs. In the following discussion, system sizes are expressed as (L,T) for brevity.

The weak scaling case is shown in more detail in Fig. 2 where it is compared with a straight line between the $N_{SPE} = 2$ point and the origin. There is close to linear scaling up to 8 SPEs, which falls away for greater numbers of SPEs. This corresponds to the regime in which some of the SPEs are on a separate chip, with a connection between the two EIBs provided over a 25.6GB/s I/O port. The almost-linear weak scaling plot indicates that the performance is limited largely by serial performance.

The (6,4) system is shown in Fig. 1 as a single datapoint, as it runs only on exactly 8 SPEs: a smaller number of SPEs would not provide sufficient LS capacity, and to use more would require a larger size in the time dimension in order to provide each SPE with at least one 3D slice. This system is of particular interest in the context of a multi-node system, along with the (4,24) system from the weak scaling series (the (4,24) system is run on 8 SPEs). The (4,24) system gives higher performance, but it is quite limited in this whole-lattice time dimensions possible: only systems with multiple-of-24 time lengths would be possible. The (6,4) system uses only around half of the available LS, but would be much more flexible in terms of simulations on different Tsizes across multiple nodes. A good compromise might be possible if some quantity of the color matrix arrays were streamed from main memory at each iteration, slightly decreasing the local store space requirement and allowing a (6,8) system on one Cell processor, at the cost of increased main-memory I/O.

Smaller systems which can be run on smaller numbers of SPEs are also shown on Fig. 1. These scale to smaller numbers of SPEs before performance starts to degrade. The only system included which can be run on the whole range of SPEs is the (2,48) size, which is small enough to run on 2 SPEs, but sufficiently long in the time dimension to run on 16 SPEs. Performance of this system falls off from 6 SPEs, although still reaches 30GFlop/s on 8 SPEs, making it a reasonably efficient choice of system size which could be useful in multi-node system for constructing power-of-two-sized global lattices.

To compare cell performance to a conventional processor, a (6,4) lattice is used to allow the conventional processor to work in L2 cache (this lattice is size of 145kB (1 spinor per site, 4 color matrices per site). A larger lattice would cause the conventional platforms to work out of main memory, while the Cell version would simply not be capable of it. However, if the Cell version were adapted to stream parts of the dataset out of main memory then it would probably remain competitive with conventional platforms due to the Cell’s relatively high memory bandwidth.

As an example of a conventional x86 processor, a 2.4GHz Xeon system was used, along with a high performance SSE3 implementation of the Dslash operation encapsulated in a library called intel_sse_wilson_dslash. This is included with the Chroma library and optionally compiled in on systems which support SSE. This provides an impressive performance of 4.5GFlop/s, demonstrating the high performance possible with this algorithm using low-level SIMD programming. However, the single precision floating point performance of the
Cell implementation presented here is almost 10 times greater than even the SSE-optimised version on the Xeon.

4. Conclusion

The code described here obtained up to 45GFlop/s performance in single precision, providing a proof-of-concept for running Lattice QCD simulations on the Cell processor. The optimisation for Cell was accomplished entirely within the C environment provided by the Cell SDK, although the extensive use of SIMD intrinsics led to less readable code. Extensions to this work could include a multi-node parallel version of this code, and a double precision version, which could be expected to achieve 20GFlop/s on the forthcoming double-precision Cell processor. However, the system sizes shown here would need to be revised if the Local Store on the SPUs was not increased accordingly for a double-precision Cell.

The Cell processor remains challenging to the scientific application programmer, demanding significant platform-specific knowledge to obtain good performance. However, for applications concentrating work in a single kernel – as Lattice QCD codes do in Dslash – the effort involved in producing a Cell-optimised version of the kernel can render an order-of-magnitude performance improvement over conventional platforms.

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