Million Atom Electronic Structure and Device Calculations on Peta-Scale Computers

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Abstract—Semiconductor devices are scaled down to the level which constituent materials are no longer considered continuous. To account for atomistic randomness, surface effects and quantum mechanical effects, an atomistic modeling approach needs to be pursued. The Nanoelectronic Modeling Tool (NEMO 3-D) has satisfied the requirement by including empirical $sp^3s^*$ and $sp^3d^5s^*$ tight binding models and considering strain to successfully simulate various semiconductor material systems. Computationally, however, NEMO 3-D needs significant improvements to utilize increasing supply of processors. This paper introduces the new modeling tool, OMEN 3-D, and discusses the major computational improvements, the 3-D domain decomposition and the multi-level parallelism. As a featured application, a full 3-D parallelized Schrödinger-Poisson solver and its application to calculate the bandstructure of $\delta$ doped phosphorus(P) layer in silicon is demonstrated. Impurity bands due to the donor ion potentials are computed.

I. INTRODUCTION

Need for Atomistic Modeling: As semiconductor structures are scaled down to deca-nano sizes the underlying material can no longer be considered continuous. The number of atoms in the active device region becomes countable in the range of 50,000 to around 1 million and their local arrangement, becomes critical in interfaces, alloys, and strained systems. An atomistic modeling approach needs to be used to capture such discreteness and quantum mechanical effects. Most experimentally relevant structures are not infinitely periodic, but are finite in size and contain contacts; such geometries call for a local orbital basis, rather than a plane wave basis which implies infinite periodicity. Furthermore we are primarily interested in stable semiconductor structures with well-established bonds which lessens or even eliminates the requirements to be able to compute the establishment of bonds with a full ab-initio methodology.

Multi-Million Atom Simulations: NEMO 3-D [1], [2] uses empirical $sp^3s^*$ and $sp^3d^5s^*$ tight binding models that have been carefully calibrated to bulk materials in the III-V [3] and Si/Ge [4], [5] material systems under various bulk strain and composition configurations. This bulk parameterization is transferred to the nanoscale under the assumption of weak charge redistributions. Weak piezo-electric effects in the In-GaAs system can be captured through strain derived charge and electrostatic potential corrections [6], [7]. Transferability of the bulk parameters to nanometer devices was demonstrated by experimentally verified multi-million atom calculations for valley splitting in Si on SiGe [8], single impurities in Si FinFETs, and InAs quantum dots in an InGaAs buffer matrix [11]. In these simulations none of the bulk parameters were modified and the nominal device dimensions were used to obtain quantitative agreement with experiment. These simulations also showed that it was essential to include millions of atoms in the simulation domain and that simplified effective mass models have led to the wrong conclusions.

Computational Cost: Multi-million atom calculations in NEMO 3-D come, however, at a typical computational price of 4-10 hours runtime on 20-64 cores on a standard cluster for a single evaluation of the eigenvalue spectrum. Inclusion of this pass electronic structure calculation into a self-consistent Poisson solution is possible, but increases the computation time up by another factor of 6-20. This drives the computational requirement into the realm of days, rather than hours. In sight of huge investments into Peta-Scale computing with availabilities of over 100,000 cores on a single supercomputer, efficient parallelism lays the goal for computational speed-up. We have been able to demonstrate NEMO 3-D scaling to 8,196 processors [12], however, such high level of scaling can only be achieved for unrealistically long essentially 1D structures, due to the 1D spatial parallel decomposition of NEMO 3-D. NEMO 3-D therefore needs significant improvements in its parallelization schemes, data handling, post-processing, and code maintainability.

OMEN 3-D: The major purpose of developing the new nanoelectronic modeling tool, OMEN 3-D, emerged out from the need for expandability in growing processor-rich environment. OMEN 3-D is equipped with more powerful parallelization engine, 3-D domain decomposition scheme and general multi-level parallelism. In addition, self-consistent charge calculations that need additional computational power are built in to deliver various kinds of scientific simulations, from impurity physics to device applications.

This paper is organized as follows. In Sections II-A and B, the parallelization schemes in OMEN 3-D and its benchmark results are presented. In Section III-C, the multi-level parallelism is briefly introduced. A Schrödinger-Poisson solver is explained in Section III-A. Section III-B contains the example of self-consistent bandstructure simulation on 2-D $\delta$-doped layer in silicon at 4K.

II. PARALLELIZATION SCHEME IN OMEN 3-D

The major feature in OMEN 3-D is its enhanced parallelization algorithm. NEMO 3-D uses a 1-D spatial decomposition scheme for parallelism. NEMO 3-D has been tested in many
supercomputers and it is proven to show close to perfect scalability, however, the maximum utilizable processors is strongly limited by its geometry in the 1D decomposition.

A. 3-D Spatial Domain Decomposition

To reduce compute times by utilization of computers in excess of 10,000 cores, a new domain decomposition scheme is introduced in OMEN 3-D. In OMEN 3-D, a device of any shape can be spatially decomposed into three dimensions and each subdomain is assigned to corresponding processor. The maximum number of processors equals to the number of unit cells in each direction. Based on the spatial information, each processor only has the list of information of the atoms in its subdomain and neighbor atoms from adjacent subdomains; no global position information is held locally, minimizing the memory consumption and making it possible to simulate large devices (Fig. 1(a)).

The major drawback for 3D parallelization is the increase of the complexity of communication by $O(n^{ND/m})$. The increased coupling among the processors may cause significant performance degradation; there is a trade-off between reducing the computational burden and increasing communication overhead. However, recent benchmark results indicated the average time consumed in the Message Passing Interface (MPI) communication is typically 5% of the total simulation time. Moreover, from the NEMO 3-D cases, it was shown that the total simulation time was not bound by communication as long as the ratio of the number of surface atoms to the total number of atoms in each subdomain is kept sufficiently small [13].

B. Benchmark Results

The strong scaling plot of the 500 Lanczos iterations using the basic 1-D parallelism for elongated systems of different number of atoms is presented in Fig. 2. This plot indicates that with minimal load of communication, OMEN 3-D shows reasonable scalability up to the structure that contains 32 million atoms with 512 processors in Ranger. However, with smaller number of atoms per subdomain, fluctuations in performance are observed as we increase the number of processors; this instability stems from the communication load being comparable to the computational operations.

The strong scaling plot in Fig. 3 examines the performance of the 3-D decomposition scheme using 500 Lanczos iterations. The structure under test is a $44 \times 44 \times 44$ silicon cube, which has 4 million atoms (80 unit cells in each direction). As with the previous strong scaling result, 1-D decomposition scheme scales linearly; the number of processors

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Fig. 1. (a) Schematic diagram of domain decomposition scheme. (b) Multi-level parallelism in OMEN 3-D.

Fig. 2. Strong scaling benchmark results of a 1D Decomposition scheme in OMEN 3-D. 500 Lanczos iterations are measured on elongated silicon structures (subfigure). The number of atoms range from 1,000 (‘1k’ in the figure) to 64 million (‘64M’).

Fig. 3. Strong scaling comparison between 1/2/3D spatial decomposition. Performance of 500 Lanczos iterations is measured on a $44 \times 44 \times 44$ silicon cube (4 million atoms). For the 2D case, the processors are assigned as $(c_x, c_y, c_z) = (16, 2^i, 1)$, $i = 0, \ldots, 4$. And for 3D case, $(c_x, c_y, c_z) = (2^i, 2^j, 2^k)$, $i, j, k = 1, 2, 3$.
can be assigned is limited to 80. On the other hand, 2-D and 3-D parallelization enables to assign more processors to the calculation, resulting in a proportional time reduction. It is measured in this example that the performance is enhanced by 13.3 times by using 16 times more processors. Therefore, by utilizing more computational resources, the 3-D decomposition scheme opens the possibility of delivering simulation results of realistic devices within significantly reduced time.

Typical NEMO 3-D simulations of InAs/GaAs Quantum Dot(QD) systems [11] not only involve electronic structure calculations but also require minimization of the total strain energy in an atomistic Valence Force Field (VFF) method [2]. This strain calculation is computationally significantly simpler than the subsequent electronic structure calculation. It therefore does not in general scale as well with increased parallelism. Here we test the VFF algorithm in 1D and 3D decomposition in OMEN 3-D (Fig. 4). The sample structure is a cylindrical InAs QD of size 20nm(D) × 5nm(H) embedded in 68 × 68 × 68 nm³ GaAs buffer, which is comprised of 13 million atoms. Again, 3-D decomposition scheme helps to scale down further to a factor of 3.5 by allocating 4 times more processors.

C. Multi-Level Parallelism

OMEN 3-D also has a programmable interface ready for multi-level parallelism as depicted in Fig. 1 (b) to achieve extra performance enhancement. In contrast to the spatial domain decomposition, where the processors are coupled to each other by MPI communication, this hierarchical parallelism solves the task independently, with different parameters assigned for each group. K-space grouping, for example, can be useful when bandstructure or charge calculations are needed. Additional bias groups can be added for simulations that may involve external electrical or magnetic fields. Depending on the application, OMEN 3-D can provide multiple levels of additional parallelism to utilize more computational resources.

III. APPLICATION

A. The Schrödinger-Poisson Solver

One of the first applications of OMEN 3-D is the self-consistent charge and potential calculation module, known as the Schrödinger-Poisson solver, which was not present in NEMO 3-D. There are three main components in the self-consistent loop:

1) **Schrödinger Equation Solver:** Solves the eigenstates of the Schrödinger equation on finite k-points based on either sp³d⁶s* tight binding or effective mass Hamiltonian using iterative eigenvalue solver, such as, (block) Lanczos or PARPACK.

2) **Charge Calculation:** Based on the eigensolutions from the Schrödinger equation, there are two different approaches to obtain the charge profile. In the case of a given Fermi level, the charge profile can be calculated simply by filling up the states. On the other hand, if the Fermi level needs to be determined by external conditions, such as charge neutrality, both the charge and the Fermi level can be determined simultaneously.

3) **Poisson Solver:** Charge is fed into the Poisson solver. The Poisson solver in OMEN 3-D also adopts 3-D parallelism and uses a finite difference method with the Aztec linear solver. The converged potential profile is iteratively obtained using Newton-Raphson’s method. The potential result is updated in the Hamiltonian and the steps are repeated until the self-consistency is achieved.

The Schrödinger-Poisson solver has been applied to a couple of physical simulations.

- **Investigation of the Charge Distribution of a Realistically Sized FinFET using the Top of the Barrier Model** [14] [15]: The non-uniform current distribution in the tri-gated devices of cross-section 65nm(H) × 25nm(W) versus the gate-voltage was successfully demonstrated.

- **Bandstructure Calculation of 1D/2D Highly P Doped silicon Structures** [16] [17]: Self-consistent bandstructure of closely positioned impurity atoms can be obtained. The self-consistent scheme applied to the impurity system is briefly introduced in the next section.

B. Example: Bandstructure of the Phosphorus δ layers in silicon

Advances in fabrication process has enabled the creation of atomic-scale devices in silicon. Using Scanning Tunneling Microscope (STM), experimentalists can fabricate phosphorus δ layers and encapsulate them in silicon [18]. This technology is significant in two fold; it is relevant in nanoelectronic device fabrication and it highlights the possibility of fabricating quantum computers. Using the self-consistent method, the bandstructure of the 2-D periodic δ layer structure with the doping density of 1/4 Mono Layer(ML) (Fig. 5) is calculated at T=4K. The bandstructure (Fig. 6) indicates that due to the potential induced by closely placed ionized donors, impurity
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