Molecular Simulations and Visualization

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Introduction

It is increasingly the case that biology, chemistry and materials science make extensive use of computational methods. The applications are diverse, spanning a range of timescales and lengthscales, from the cellular level (e.g. in systems biology) all the way down to detailed atomistic simulations of molecular assemblies, materials or small molecules. The 2013 Nobel Prize in Chemistry awarded to Martin Karplus, Michael Levitt and Arieh Warshel for the development of multiscale models for complex chemical systems is an indication for the achievements of such simulations. Molecular simulations and visualization offer fertile territory where research in Human-Computer Interaction and Virtual Reality may interact with and provide substantial benefit to computational molecular sciences. One of the principal challenges of this research area is that it is inherently cross-disciplinary and therefore requires deep exchanges beyond the boundaries of each discipline. In what follows, we outline progress in this emerging field, and offer a glimpse of potential directions. The discussion that follows is broken down into four interconnected topics. The first topic focuses on the use of virtual and augmented reality in the context of immersive molecular simulations. Progress here requires advanced visualization and visual analytic techniques, as well as the ability to harness new developments in high performance computing (e.g., general-purpose-GPUs (GP-GPUs), clouds and more), which respectively form the second and third topics in our discussion. The final topic discusses a relatively young area where progress relies on fusing the previous three topics – namely, the development of applications and serious games: from docking to model building.

Virtual and augmented reality and immersive molecular simulations

Setting up, running, and analyzing a molecular simulation is generally a lengthy process that requires considerable user patience and expertise. In principle, interactive simulations provide a much lower latency approach for manipulating and exploring molecular structures [¹]. Such methods provide a virtual or augmented reality framework for immersing the user within the simulation, with the aim that molecular interaction becomes as intuitive as possible [²]. One of the earliest and perhaps most obvious methods for facilitating a more immersive molecular interaction experience involves the sense of touch. Indeed, the use of physical models has a long history within both chemistry and biochemistry, perhaps most famously captured by Watson and Crick's physical model of DNA. With the advent of computing and robotics, touch is a sensory channel that continues to receive a great deal of attention, mostly through the use of haptic interaction strategies [³,⁴,⁵]. Touch is also exploited in systems that utilize tangible physical models [⁶] that may be augmented virtually [⁷] and which have shown

promise in enhancing student learning [⁸]. Beyond touch, a variety of sensory channels related to visual and audio feedback may be used to enhance the immersive effect, and preliminary applications of such integrated methods have occurred in the context of docking problems [⁹]. The design and utilization of efficient and effective strategies for interaction with a molecular simulation relies on careful consideration of methods developed within the field of human-computer interaction.

In general, an interactive molecular simulation framework requires several different ingredients: (1) a mechanism for representing the molecular system – usually involving a screen-based visual display and/or a physical object; (2) a mechanism for interaction with the simulation - typically through dedicated peripherals like haptic devices, cameras, a mouse, etc.; (3) animation of the simulation using a physics-based simulation engine; and (4) a pipeline which allows low latency coupling of these elements - often utilizing some sort of network protocol. In contrast to many fields of scientific computation, latency is an important consideration for interactive systems, and important bottlenecks need to be ameliorated -i.e., the simulations need to be fast enough to run at an interactive speed, and the visualization tools need to be sufficiently elaborate and flexible so that the user can tackle both very complex and heterogeneous data. Interactive simulations are presently supported in many software packages, including NAMD [¹⁰], LAMMPS [¹¹], HOOMD [¹²], ProtoMol [¹³] and OPEP/HireRNA [¹⁴]. As far as the simulation engines are concerned, one of the principal concerns is that of performance: for a reasonable interactive experience, animation of the molecular system needs to occur with refresh rates of at least 24 Hz. Parallelisation therefore plays an important role in interactive simulation, because it can give substantial increases in the performance of molecular dynamics simulation engines. Stream architectures like GP-GPUs (as described below) provide one way to exploit parallelisation and achieve performance increases. [¹²,¹⁵] Compared to standard single-core environments, data transfer tends to present a significant bottleneck in massively parallel environments, resulting in a number of challenges – e.g., extracting atomic positions without degrading performance in a stream parallelized software package is not trivial. More generally, MD simulations can produce a large quantity of simulation data, and transfer of this data may generate bottlenecks for coupling with visualization and interaction modules. Interactive all-atom simulations have recently been reported for systems up to 2 million atoms $[1^{6}]$. For the exploration of large systems, coarse graining offers another well-balanced alternative to speed-up the physical engine driving an immersive simulation experiment $[^{17}]$.

There are a wide variety of potential applications for interactive molecular simulation frameworks, including structural modelling, conformational searching, and interpretation of the mechanisms that drive function in complex biological models. Drug design is another active area in which new interactive simulation strategies are under investigation (including haptics interaction, virtual reality, and 3D printing [¹⁸]), in part motivated by the very high cost of bringing new medicines to market. Interactive simulations have been used to facilitate forms of nano-manipulation and even to prototype and design nano-robots [¹⁹,²⁰]. Interactive simulation frameworks already offer considerable potential for providing microscopic insight into experiments; however, an even closer linking with experiment is likely to emerge in the near future. For example, a multitude of biophysical techniques (SAXS, cryo-EM, FRET, mass spectrometry, etc.) nowadays routinely generate a huge amount of biological data, opening up the possibility of combining all the data to build increasingly accurate models [²¹,²²,²³]. It will soon be possible to extend immersive approaches to explore not only simulation data, but experimental information as well (e.g. from NMR spectroscopy [²⁴]) and to subsequently build models from such data under direct human supervision [²⁵].

application of interactive molecular simulation involves re-constituting molecular assemblies from cryo-EM data [²⁶]. These sorts of applications open a whole new range of collaborative opportunities and questions [^{27, 28, 29}].

Advanced visualization and visual analytics

One of the cornerstones of modern molecular simulation concerns the visual representation of the structure of a molecule and its properties. Visual representation is particularly important in guiding the manner in which scientists think about atomic and molecular structure, which is partly a result of the fact that our human perception requires some form of augmentation in order to 'see' this world [³⁰]. Attempts to conceptualize and visualize molecules reach far back in the history of chemistry. In terms of three-dimensional molecular structures, a notable milestone along this path goes back to the early 40s when Roger Hayward depicted the arrangement of atomic assemblies in collaboration with Linus Pauling in both a scientifically accurate and aesthetically pleasing way [³¹]. Since this era (and in particular recently), technical progress has considerably improved our ability to visualize the molecular world. Nowadays, molecular graphics are ubiquitous and every scientist can display the structure of a biomolecule on his/her personal computer $[^{32}, ^{33}]$ or tablet device $[^{34}]$. To ensure that the enormous quantity of information contained in molecular simulation data (interactive or not) furnishes maximum insight into microscopic phenomena, the investigation of new visualization and visual analysis methods is an area of active research. One of the primary focuses of this emerging area concerns the development of new ways to understand and rapidly process the radically expanding deluge of data which molecular simulations are capable of generating. Visualization assists in grasping the complexity of these data and identifying emerging properties.

It is now possible to interactively visualise very large molecular assemblies, and new developments (including the use of GPU programming) are driving performance gains and opening up new possibilities for visualization [35]. Nowadays, millions of atoms and their bonds can be depicted interactively [36 , 37 , 38], with considerable speed-ups in secondary structure representation [³⁹]. Calculating molecular surfaces on-the-fly is more demanding $[{}^{40}, {}^{41}, {}^{42}, {}^{43}]$, but realistic rendering including the effects of lighting and ambient occlusion $[{}^{44}]$, reaches real-time refresh rates [⁴⁵]. Impressive progress has been achieved with interactive raytracing systems the GPU (http://www.molecularof molecular on visualization.com/#!home/mainPage). Using ray-casted instancing, even whole-cell simulation data may be visualized smoothly [⁴⁶] on stereoscopic displays, [⁴⁷] allowing the reconstruction of 3D cellular complexes built from proteins and DNA molecules [⁴⁸].

The widespread availability of molecular simulation packages means that molecular visualization must have grown accustomed to depicting time-dependent dynamics, calling for on-the-fly visualization of atomic and molecular interactions (e.g., hydrogen bonds) [³⁷], molecular properties such as helix bending [⁴⁹], or the dynamics of molecular paths and cavities [^{50,51,52}]. Visual analytics (http://www.visual-analytics.eu) have great potential to aid in understanding the increasing number of simulation datasets, and have been applied in a few cases [^{53,54}]. Simplifying large quantities of complex data like that produced by MD simulations may be achieved by appropriate abstractions. In this context, techniques from scientific and medical illustration are helpful and have found their way into the visualization of chemical structures [^{55,56,57}]. A stimulating recent example is the continuous abstraction of a molecular illustration [⁵⁸] to yield a continuum of molecular depictions. Another challenge

that arises in particular for the visualization of molecular simulations concerns the depiction of molecular flexibility [⁵⁹]. In fact, chemical reactions themselves are difficult to render for many visualization tools. More generally, the visualization of dynamic molecular interaction networks is a very active field of research [⁶⁰,⁶¹,⁶²], but is beyond the scope of this short introduction.

The ubiquity of molecular images and associated visualization tools is in part a consequence of the fact that it has been benefitted from other high-growth economic areas. For example, tools that are traditionally dedicated to areas such as the video game industry or the film industry may be used for molecular visualization $[^{63}, ^{64}, ^{65}]$ or animation $[^{66}, ^{67}, ^{68}]$. The availability of such tools has enabled the use of molecular visualization in collaborative structural biology, for example using TV-based $[^{69}]$ or web-based $[^{70}]$ solutions. A similar cross-fertilization is observed for GPUs, originally used in the consumer graphics market and nowadays omnipresent in high performance computing and scientific visualization.

Computing power revolution and new algorithms: GP-GPUs, clouds and more

The field of molecular simulation and visualisation intrinsically depends on highperformance computing (HPC) to ensure the underlying calculations can be carried out in real time and on a broad range of hardware including commodity computers. In this context, general-purpose-GPUs [⁷¹], cloud computing, [⁷²] and many-core architectures [⁷³] are finding their way into the molecular simulation community. Multi-core architectures are evolving quickly, with massive-parallelism and massive-threading available on machines like the 1.3 million thread Blue Waters supercomputer. Bespoke architecture development, like that available on the Anton machine, is similarly allowing researchers to push the boundaries of simulation [⁷⁴] and new techniques based on cloud-based methods and ultrafast highperformance networking are just around the corner. These and likely future developments in HPC are making massively parallel computations viable, and are stimulating innovation across hardware, software, and hardware/software integration much of which is aimed at tackling the main challenges of molecular simulation: the size of systems which can be simulated, the time-scales which it is possible to simulate, the ability to sample large regions of molecular phase space, and the rigor of the underlying physics within the models. Lane et al. ⁷⁵] have touched on many of these aspects in the context of protein folding.

The advent of programmable GPUs [⁷⁶] using high-level languages like C, in conjunction with the NVIDIA CUDA (Compute Unified Development Architecture) tools, OpenCL and other frameworks has been instrumental in porting software and developing new algorithms. The rise of GPUs offers another example of how advances in high-growth consumer markets (namely video gaming) has been exploited for the purposes of scientific simulation. As a result of the power of GPUs, and the fact that they are relatively inexpensive, much academic and commercial molecular dynamics (MD) software (e.g., GROMACS and AMBER) [⁷⁷] has been GPU accelerated. The enhancement in speed can vary, in part due to the specific algorithms used, and also as result of the particular GPU hardware architecture on which the code is run, both of which lead to different scalability and execution time. Adding many-body terms to potentials used in classical simulations is a case where the computational cost has been mitigated through exploitation of new hardware, by the development of a shared-memory force-decomposition algorithm [⁷⁸]. Calculations using Reax-FF, which is a reactive force field, have been accelerated using GPUs [⁷⁹]. Aspects of reliability and reproducibility have been studied in the context of error correcting code [⁸⁰].

For quantum chemistry, software adoption of GPUs has been slower than for MD simulations, but building on initial work [^{81,82,83}] many electronic structure packages now have GPU-enabled codes, and there is significant interest in utilizing fast quantum chemical methods, for example, to investigate reaction mechanisms [⁸⁴]. Recent work has investigated GPU acceleration in a range of contexts, applied to: (1) double precision matrix multiply operations within legacy quantum chemistry codes [⁸⁵]; (2) ONETEP, a linearly-scaling plane wave density functional theory (DFT) code; [⁸⁶] (3) BigDFT, a hybrid DFT code based on Daubechies wavelets; [⁸⁷] (4) VASP, GPU-accelerated electronic structure calculations; [⁸⁸] (5) real-space DFT implementations within the Octopus code; [⁸⁹] and (6) semiempirical methods [⁹⁰]. Very recently, Sisto et al. have outlined fragment-based quantum chemical methods which rely on both distributed and shared memory GPU parallelism to carry out very large excited state TDDFT calculations using the TeraChem software framework [⁹¹].

Cloud computing [92,93,94,95] is a relatively recent approach to molecular simulation that builds on distributed computing approaches like FightAIDS@home, SETI@home, and folding@home. It has already been used for artistic applications at the intersection of art, science and culture, notably in relation to molecular aesthetics. Distinct from other highperformance and distributed paradigms, it provides large-scale compute infrastructure on demand. In many respects, cloud-based approaches are still in their infancy, but are attracting growing attention. For applications of molecular simulation and modelling, cloud computing can offer large-scale data and compute capability for a short 'burst' phase. Cloud computing provides another example wherein molecular simulation benefits from exploiting approaches which have applications in other sectors: for example, cloud-based computing has appeal to small and medium biotech start-ups where continuous in-house HPC facilities would be under-utilised. Embarrassingly parallel tasks, like the generation of combinatorial databases, virtual screening of millions of compounds, and the analysis of the huge genome datasets, are well suited to existing cloud provisions. A workflow system called AutoDockCloud [⁹⁶] enables distributed screening on a cloud platform using the molecular docking program AutoDock. For applications with greater demands for inter-processor communications, scalability is a key issue. A plugin [97] for the popular VMD software [98] (a front-end for NAMD [10] allows one to (1) create a cloud-compute cluster on Amazon EC2; (2) submit a parallel NAMD job; (3) transfer the results back for subsequent post-processing; and (4) shutdown and terminate the compute cluster on Amazon EC2. These and other case studies of molecular modelling using cloud computing have been reviewed by Ebejer et al. [⁷²].

Crowd-Sourcing and serious games: from docking to protein folding

Molecular simulation, like many areas of computational science, involves a tradeoff between user control and automation. Users usually have a deeper understanding and context for the problem at hand, but limited speed and memory. Computational systems, on the other hand, excel in memory and speed, but are limited when it comes to understanding and context. Even with the tremendous advances in computation discussed in the previous section, it is likely the case that there will always be a limit to the size and accuracy of models that can be built for a particular system, and therefore some level of human understanding will always be required. It is therefore of fundamental interest to consider radically new approaches to molecular modelling - i.e., utilizing paradigms that do not rely exclusively on ever-faster computational frameworks.

Very recently, there has been a great deal of interest directed at investigating whether human intuition and problem solving skills can be effectively mobilized (usually via the Internet) as a new resource for solving research questions [⁹⁹]. The interest in these solutions is such that participants may be stimulated by the prospect of being remunerated $[^{100}]$. Success in this area requires that the research approach or proposition is cast in a way that is sufficiently engaging, entertaining, or educational. Along these lines, the Defense Advanced Research Agency (DARPA) recently developed a challenge to see how quickly it is possible to involve a large number of people to fulfil a particular task [¹⁰¹]. Such 'crowd-sourced' research approaches [¹⁰²] have received increasing attention. For example, one particularly successful example is the GalaxyZoo $[^{103}]$ project, which transforms a potentially mundane, but difficult computer vision task (classifying images of galaxies) into an attractive challenge. When it comes to solving scientific research problems, collective and intrinsic motivation can marshal large communities of volunteers. This requires a high-level of visibility, which social media and modern communication technologies can successfully facilitate. Once a volunteer community is established, strategies and structures must be in place to maintain the ongoing engagement of the community. In many cases, crowd-sourced scientific computing paradigms raise interesting questions related to data analysis and data integrity.

Crowd-sourced approaches to research can generate useful insight owing to user intuition as a solution to cope with complex data and unveil emerging properties: a striking example is the game Foldit [104 , 105]. This project presents protein folding as a sort of three-dimensional jigsaw puzzle, where players are invited to shake and wiggle the three-dimensional structure of proteins in order to find the most stable conformations. Since May 2008, when the first beta version of this game was released, the project has gathered a large community. In some cases Foldit players have been able to find optimal structures that automated search strategies failed to sample. Players do not necessarily require significant knowledge of biology to play the game and to find stable protein configurations. It is more a matter of spatial representation in three dimensions, as well as collaboration between players. The first 'levels' of the Foldit game are designed to train the players in order to accomplish increasingly complicated tasks. Interactions among players have led to remarkable results from a biological point of view [$^{106}_{107}$, $^{108}_{108}$] but also to develop collaboratively new algorithms to solve a particular problem [$^{109}_{109}$].

Not only do interactive and video game interfaces offer the potential for crowd-sourced research studies; they also offer an engaging medium for scientific education, helping students of all ages learn scientific principles and knowledge. As a consequence, educational games are flourishing. For example, the Spectral Game [¹¹⁰] seeks to teach quite advanced concepts in spectroscopy, specifically proton nuclear magnetic resonance (NMR). In addition to meeting specific targets, educational games and interactive molecular simulation platforms (like the distributed computing projects discussed above [¹¹¹]) have a more general effect – i.e., they engage the public and thereby increase public awareness and understanding of scientific problems. New channels for engaging the public with scientific ideas are also emerging in less traditional venues – i.e., on the frontiers of aesthetic imagination and scientific visualization. As art moves increasingly toward digital mediums, artists have become fascinated with the glimpse into the invisible atomic world provided by molecular simulations and visualizations, to the extent that it has inspired new forms of artistic expression and aesthetic content [³⁰].

Conclusion

Molecular simulation and visualisation represent a vibrant melting pot of many scientific disciplines that both benefits from and drives significant progress across a range of fields. New hardware architectures, new software algorithms, and new technological developments drive this evolution and herald an exciting era of increasingly sophisticated and perhaps unconventional molecular simulations. The potential for these new simulation frameworks is extremely exciting: they will allow us to obtain unprecedented new research insights, develop new ways for interacting with and imagining the microscopic world, drive progress in HCI and computer science, and ultimately have profound effects beyond the scientific realm within the broader culture.

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