Teaching Chemistry with Computers

Hua-Jun Fan, Joshua Heads, Daniel Tran, and Nnenna Elechi

Abstract—With a new generation of undergraduate students who are savvy with technologies and computers, coupled with the advancements in computer hardware and software development, the learning curve of computational chemistry is diminishing. We demonstrate various modules that can be used not only to illustrate the difficult concepts in chemistry education, but also to provide the hands-on experience that allows students to generate their own data and conclusions.

Index Terms—Computational chemistry, molecular modeling, undergraduate chemistry curriculum, American chemical society.

I. INTRODUCTION

The new generation of undergraduate students is nothing like the type of students we used to teach [1]. This new generation wants to learn things differently and quickly, and tends to have less patient to read books, though we believe that’s the only way to learn the new information. To make things worse, some of these kids will never open their books before the exams, sometimes not even before the end of semester. They would search answers and concepts on Google, regardless of the accuracy and correctness of the answer. What drives some of us crazy is when some of these students would simply copy and paste online answers and claim their credit, which we call plagiarism. Chemistry is no different than any other science subjects that has encountered this very same problem.

According to a survey conducted in 1995 [2], many students describe chemistry as one of the toughest and most boring science courses they must pass, despite the fact that chemistry, a science central to technology and engineering fields in many ways, is an easy subject to apply to real life [3]. The examples and applications of chemistry are abundant and unavoidable in everyday life. For example, an understanding of fundamental chemistry concepts are required to solve the oncoming energy crisis, to develop environmentally friendly methods of production and waste management, to detect biological weapons in real time, to develop better pharmaceutical drugs, to solve environmental problems, and to design newer functional nano-materials. Therefore, chemistry along with other science courses such as physics and math usually are weaved into the university core courses that almost all incoming freshmen will have to take and pass.

Depending on the major and classifications, most campuses will have a chemistry course for students majoring in non-science related subjects such as nursing, technology or engineering [4] and there are separate courses set up just for the chemistry, biology, and other science majors. Because the needs and the purposes of these courses are different at various campuses, there is no universal curriculum for these courses. However, all of these chemistry courses require a significant amount of foundational knowledge of electronic structures, electron configurations. After they’ve learned these basic fundamentals, we then demand the student to apply and correlate between the structure and property, and finally predict the properties of new materials. The job market demands our students to at least have some type of exposure. For example, the new (revised) undergraduate curricular guidelines from American Chemical Society Committee on Professional Training (ACS-CPT) suggest the foundation-level chemistry curriculum should be reorganized into three sequences: structure, reactivity, and quantization. It is believed that the new reorganization would allow students to more quickly appreciate the breadth of the field than the traditional domains.

Ever since the National Science Foundation (NSF) reformed the Chemistry curriculum with the Chemical Bond Approach (CBA) and Chemical Education Material Study (CHEM Study) in the late 1950s, and students were learning chemistry through hands-on laboratory procedures that required problem solving, and a stronger technical training in chemical principles across the board [5]. With the dilemmas and the new challenge brought upon us by this new generation of learners, they have also brought new opportunities. For example, most of them now have tons of electronic gadgets at their disposal and most of time they are online for social or entertainment purposes. There have been some researchers and educators who embrace electronic teaching such as distance learning, podcast, mini-video streaming, along with clickers, traditional power points, and transparencies. In this paper we will explore computational chemistry in undergraduate education. In particular the molecular modeling tools that are used to 1) help students understand difficult concepts such as electron configuration and structural modification, in general chemistry setting, and 2) as a research tool for advanced chemistry courses or scientific research.

II. APPROACHES

There are two different types of students who take chemistry: the science majors and non-science majors. In general, the science majors usually find it easier and would adapt to, if not already being trained on, the traditional conceptual approach of learning chemistry, which often is taught in the form of rigorous facts and principles and very much in an abstract manner. On the other hand, the non science majors are still in need of developing some types of
scientific (aka chemical) literacy, which would then allow them to understand the principles, and engage in an interactive way to grasp the materials and possibly apply those principles to better understand the phenomenon around them from environmental effects to nanotechnology [6]. It is not surprising that the non-science majors find this traditional approach to chemical education difficult and boring, and are struggling to understand the relevance of conceptual chemistry. Most of these students will lose their interests in science, particularly in chemistry within the first year [7] because of the embedded requirement in core curriculum.

III. METHODOLOGY

In this study, we will use computational technology to expound upon various chemistry situations. As examples, we will first use it to describe the kinetics of substitution reactions. Then we will use it to explain to students the importance and effectiveness of sunscreen. Finally, we will describe how computational chemistry can be used not just in the methods listed above but in other aspects of chemistry.

By using this tactic, it allows students to produce the data quickly and draw their own conclusions as they learn from the textbook. The approach used here is to adopt the current graphic user interface (GUI) such as WebMO, Spartan, or GaussView in order to facilitate the steep learning curve of basis sets, functionals, methods and software. So the students can focus on learning the chemistry and concepts, rather than on the how-to’s and set-up’s. For the sake of time, semi-empirical methods such as PM3 and ZINDO [7] were used. The software used in this study is GaussView and Gaussian 09 [8]. The geometries of the compounds in the developed modules were optimized under B3LYP functional [9] with Pople’s 6-31G* basis set [10].

IV. RESULTS AND DISCUSSION

One of the important and vital tasks as a chemistry educator is how to move away from the perception that chemistry is difficult and boring. This perception was developed during the early stage of Chemistry and physics around 1920s when more fundamental discoveries helped form the theoretical principles. These discoveries allowed chemistry to become more rigorous and analytical, which also made the field highly submerged in tradition [11]. As such, the material that needed to be covered in general chemistry exploded over the years. To make things worse, the new materials were simply augmented and added without evaluating their relationship to old principles. A review by Lloyd in 1992 pointed out that the typical general chemistry textbook changed from a small 5 inch x 8 inch book into a 1000+ page, 8 inch x 10 inch encyclopedia book that averaged 6 lbs [12].

Clearly there is a need to modernize how chemistry is taught to our new generation of students. There are many published efforts and research that has been invested in endeavors to assist students’ learning, such as MIT’s OpenCourseWare [13], Process Oriented Guided Inquiry Learning (POGIL) [14], Peer-led Team Learning (PLTL) [15] and context-led approaches (CLA) [16]. As their names suggest, all these approaches actively involve the students either on problem-based or real life scenario-based self-instruction and self-teaching. The results would have been somewhat successful if the approach was well planned and implemented [17]. While our approach in this study also adopted student-centered engagement, we employed computational chemistry and molecular modeling as a tool to facilitate the student learning.

The computer technology and software development has enabled modeling tools to be used on a par with experimental methods as a legitimate and practical means for exploring chemistry. Molecular modeling can offer major benefits as a tool for exploration such as studying a compound that is difficult to synthesize in a laboratory setting. The cost of making this compound on screen is essentially just the price of the software itself and the creator’s imagination. However, one must be very cautious of such actions as the traditional wisdom says “Garbage in, garbage out.” Therefore, the guidance from a more experienced user and instructor is essential in teaching the “right” chemistry. That way, students will not be taught improperly. Compared to the cost of synthesis, purification and characterization costs, modeling tools providescrucial information on geometries, 3-D rendering, volumes, contact areas, symmetries, reaction mechanisms, and energy profiles such as activation energies for kinetics and thermodynamic parameters such as enthalpy, entropy and Gibbs free energies.

Like any discipline, good science instruction should start with well drafted learning objectives and learning outcomes as one of the essential tools for the success of the course and a guide for students’ learning. Literatures show that learning objectives and learning outcomes follow more than two dozen taxonomies that have been developed to define the domains of learning, development, and cognition [18]. However, most of them were all based on Bloom’s classic Taxonomy developed in 1956 [19] as shown in Fig. 1. While Bloom’s classic Taxonomy of Educational Objectives were defined by six hierarchical levels of cognitive processing, (knowledge, understand, apply, analyze, synthesize and create), a more modern version of Bloom’s Taxonomy use a non-hierarchical definition of learning (knowledge, comprehension, application, analysis, synthesis, and evaluation).

![Bloom’s Classic Taxonomy](image)

**Fig. 1.** Bloom’s classic Taxonomy developed in 1956.

Though the words of these versions are similar, the approach to assist students’ learning is quite different. The classic hierarchical approach assumes students’ progress gradually from bottom up, while the non-hierarchical
approach assumes students’ learning at all levels and all aspects. These different learning skills were developed concurrently, therefore training of these skills need to be applied correspondingly. There are even other types of taxonomy such as foundational knowledge, application, integration, human dimension, caring and learning how to learn [20]. Computational chemistry and modeling tools actually enable students to apply, analyze and synthesize the chemical concepts embedded in the question through the data generating, collecting, and analyzing.

Here, we will demonstrate some of the modeling modules adopted [21] and developed to illustrate how computational chemistry can effectively explain and visualize the difficult and abstract chemical concepts.

A. Kinetics of Substitution Reactions

The reaction pathway and its theories are something that is usually demonstrated via diagrams such as Fig. 2.

![Energy diagram of a reaction coordinate for a simple one-step reaction.](image)

According to the collision and transition state theory, the reactant molecules must effectively collide (geometry-wise and energy-wise) in order for the reactants to enter a transition state (TS) before they can form the products. This collision effectiveness is the key to the activation energy as depicted as $\Delta E_a$ in Fig. 2, which is the minimum energy that must be supplied by the collisions in order for a reaction to occur. The energy difference between the reactants and products determines the endothermic or exothermic nature of the reaction (See Fig. 3). Students have a hard time picturing the transition states and the collision. Therefore, a carefully designed reaction such as a one-step nucleophilic substitution reaction (SN2) of bromide with methyl chloride, could be used to illustrate the concepts. The steps implemented are:

1) Prepare the reactants CH\textsubscript{3}Br + Cl\textsuperscript{-} and the products CH\textsubscript{3}Cl + Br\textsuperscript{-}
2) By fixing C-Br bond, scan the bond distance between C-Cl from 1.7 to 4.0 Å
3) Swap Br-Cl position, then fix C-Cl bond distance to the lowest energy point in above scan, scan the C-Br bond distance from 1.7-4.0 Å.
4) Deduce the possible geometry of transition state based on above steps
5) Use Scan function of Gaussian to set up automatic 2-D scan of C-Cl and C-Br distance and plot the geometry of transition state to be compared with the results above steps
6) Plot the relative energy of the reactants CH\textsubscript{3}Br + Cl\textsuperscript{-} the products CH\textsubscript{3}Cl + Br\textsuperscript{-}, and transition state

This process actually can be used to serve three folds of teaching: first, to provide a visualization of the geometries of reactants, products and especially the transition state motion. For example, the frequency of the transition state can yield the concerted motion of leaving group and entering group, such motion is extremely valuable to demonstrate TS.

![Visualization of the reactants and products.](image)

Second, providing the understand of the relationship among the Lewis dot structure, electro negativities of Br and Cl, geometry and dipole moments. As shown below in Table 1, the calculated geometric parameters are carbon-halide bond distance (Å), dipole moment (Debye), and atomic polar tensors charge and population analysis (e\textsuperscript{-}). Students can easily draw the correlation between these calculated parameters to the electro negativities and geometries of reactants and products. While the explanation provided by most textbooks is readily available, this process allows students to produce the quality data, formulate and synthesize their own conclusions.

<table>
<thead>
<tr>
<th></th>
<th>Electro negativity</th>
<th>Dipole</th>
<th>C-X bond</th>
<th>Charge on X</th>
</tr>
</thead>
<tbody>
<tr>
<td>CH\textsubscript{3}Br</td>
<td>2.96</td>
<td>2.132</td>
<td>2.011Å</td>
<td>-0.232</td>
</tr>
<tr>
<td>CH\textsubscript{3}Cl</td>
<td>3.16</td>
<td>2.477</td>
<td>1.877Å</td>
<td>-0.068</td>
</tr>
</tbody>
</table>

Third, students can then search and deduce the TS geometry using steps 2 and 3, while the software generates and pinpoints the TS on the potential energy surface (PES) of reaction pathway in step 5. A PES scan along C-Br and C-Cl bond direction provides students the approximate location of the transition state as shown in Fig. 4. This process could allow students to see the principle and foundations behind the software in search of the transition state as well how the software identifies and evaluates the kinetics of a reaction.

![The potential energy surface (PES) scan along C-Br and C-Cl distance.](image)

B. Understanding the Effectiveness of Sunscreen

Computational chemistry can not only tackle the difficult concepts as demonstrated above, but also investigate relevant cases such as the effectiveness of sunscreen.

These types of exercises are similar to the context-led approaches, which use real life scenarios to understand...
chemistry principles. For example, the purpose of sunscreen is to block out the harmful ultraviolet (UV) lights such as UVA (320-400 nm), UVB (290-320 nm) and UVC (100-290 nm). The commonly available sunscreen chemicals utilize (1) physical blockers such as zinc oxide and titanium dioxide, which gives the white color, and the conjugated \( \pi \)-system molecules to absorb (aka block) UV lights. Because the atmospheric gases and ozone can absorb most of UVC lights, the main function of sunscreen is to absorb UVA and UVB lights. This study can make a role-play by asking students to design a new sunscreen mixture for a new line of sunscreen products.

The type of conjugated systems such as those chemical compounds in Fig. 5 can be used to model the UV-Vis spectra as those measured by UV spectrophotometer. Of course students can use any other compounds in any chemical catalog. The main method used in this modeling is a semi-empirical quantum chemistry named Zerner's Intermediate Neglect of Differential Overlap (ZINDO) [22]. ZINDO can produce the excited states and UV-Vis spectra of these compounds. The predicted UV-Vis of these eight compounds is shown in Fig. 6. By examining the main absorption peak at various wavelengths, students can then identify which one(s) are the best combinations for sunscreen.

**C. Other Modules**

There are many online resources such as a tutorial for the software available for computational chemistry and molecular modeling in chemistry instruction. Especially in organic and physical chemistry, there are many examples in the textbook can be modeled and compared. Table II lists the example list of modules developed.

**TABLE II: THE EXAMPLE LIST OF MODULES DEVELOPED**

<table>
<thead>
<tr>
<th>Title</th>
<th>Concepts explained</th>
</tr>
</thead>
<tbody>
<tr>
<td>Identify isomer equilibrium</td>
<td>Chirality and R/S designation, Equilibrium constant.</td>
</tr>
<tr>
<td>Comparison between Benzene and 1,3,5-cyclohexatriene</td>
<td>Enthalpy, resonance stabilization, enthalpy of formation, delocalization effect</td>
</tr>
<tr>
<td>Vibrational analysis of functional groups</td>
<td>Infrared spectra, electron donating and electron withdrawing groups</td>
</tr>
<tr>
<td>Rotational barrier investigation</td>
<td>Resonance structure, delocalization of ( \sigma )-bond, effective barrier</td>
</tr>
<tr>
<td>Polarieties of molecules</td>
<td>Dipole moments, Lewis dot structure, VSEPR model, electronegativity</td>
</tr>
<tr>
<td>IR spectra vs molecular geometry</td>
<td>Group theory, point group, irreducible representation, infrared and Raman spectra, frequencies</td>
</tr>
<tr>
<td>Molecular orbitals</td>
<td>Molecular orbital theory, bonding and anti-bonding orbitals, symmetry</td>
</tr>
<tr>
<td>Octal rule and 18-e rule</td>
<td>Electron configuration, molecular orbital theory</td>
</tr>
<tr>
<td>Ligand field strength</td>
<td>Ligand Field Theory, highest occupied molecular orbital and lowest unoccupied molecular orbital</td>
</tr>
</tbody>
</table>

**V. CONCLUSIONS**

Computational chemistry and molecular modeling tools are becoming available and accessible to students. The calculation ability also becomes more and more powerful due to advancements made in the hardware design and software development. In this study we demonstrated various approaches that can be utilized to exemplify the difficult concepts found within the chemistry education, affording the students the necessary hands-on experiences that are essential in allowing them to generate their own opinion and conclusions.

**REFERENCES**


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Hua-Jun Fan was born in Zhejiang Province, China in 1970. He graduated from University of Arizona as PhD in chemistry in 1999.

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