Relationship between Atomic Structure, Composition, and Dielectric Constant in Zr–SiO₂ Glasses

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ABSTRACT: Computational methods, or computer-aided material design (CAMD), used for the analysis and design of materials have a relatively long history. However, the applicability of CAMD has been limited by the scales of computational resources generally available in the past. The surge in computational power seen in recent years is enabling the applicability of CAMD to unprecedented levels. Here, we focus on the CAMD for materials critical for the continued advancement of the complementary metal oxide semiconductor (CMOS) semiconductor technology. In particular, we apply CAMD to the engineering of high-permittivity dielectric materials. We developed a Reax force field that includes Si, O, Zr, and H. We used this force field in a series of simulations to compute the static dielectric constant of silica glasses for low Zr concentration using a classical molecular dynamics approach. Our results are compared against experimental values. Not only does our work reveal numerical estimations on ZrO₂-doped silica dielectrics, it also provides a foundation and demonstration of how CAMD can enable the engineering of materials of critical importance for advanced CMOS technology nodes.

INTRODUCTION

Complementary metal oxide semiconductor (CMOS) device scaling has been driving the need for high relative permittivity (high-k) dielectrics. During fabrication, high-k dielectrics are typically either directly deposited on top of the silicon channel (one layer oxide) or on an intermediate layer of SiO₂ with a thickness of approximately 1 nm.¹ A dielectric material such as Hf/ZrO₂ was considered to replace the otherwise commonly used SiO₂.² Of the two, ZrO₂ found fewer applications in advanced CMOS technologies³−⁵ and is therefore less studied.

This article reports on our work that used molecular dynamics (MD)-based computer-aided material design (CAMD) methods to predict the dielectric constant of amorphous solids. In particular, we use classical MD to model bulk amorphous (ZrO₂)x(SiO₂)1−x, where the ZrO₂ content can be tuned to achieve a certain desired dielectric constant. CAMD constitutes a set of alternative approaches for material discovery and characterization which can potentially reduce the associated material discovery costs via prescreening and redirecting the physical lab efforts. Different CAMD algorithms have been devised and used to date. These different algorithms are generally based on numerical simulations that, in turn, are based on different levels of abstractions of the physical material at hand.⁶,⁷ Among these different algorithms, those generally referred to as ab initio methods are known to be the most accurate and have successfully been applied for material discovery purposes. Unfortunately, while ab initio methods may be the most accurate, they tend to be computationally prohibitive for handling any “at scale” molecular system. Methods that have been devised to overcome the latter limitations by using simpler models include classical MD methods.

In our specific application of MD-based CAMD, we essentially adapted and customized an approach previously used in various contexts by other groups.⁸−¹² This approach consists of parameterizing and optimizing classical forcefields that are subsequently used in MD simulations to determine particular characteristics of new or existing materials. Our CAMD algorithm can be broken down into two major procedural steps: a Force field Preparation step followed by a MD simulation step. The force field preparation consists of the following:

(A) Ab initio simulations: Quantum chemistry methods are used to perform calculations on a set of molecular structures to characterize them in terms their ground state energy, structure, partial charges, and so forth.
Table 1. General Structural Characteristics of (ZrO$_2$)$_x$(SiO$_2$)$_{1-x}$

| structure | Si–O (Å) | Zr–O (Å) | O–Si–O (deg) | O–Zr–O (deg) | Si–O–Zr (deg) | Si–O–Si (deg) | CN | density (g/cm$^3$) |
|-----------|---------|---------|-------------|-------------|--------------|--------------|-----|------------------|
| x = 0.00  | 1.62    |         |             |             |              |              | 3.98| 2.20             |
| x = 0.02  | 1.63    | 1.88    | 109.09      | 108.44      | 142.23       | 137.69       | 4.08| 2.25             |
| x = 0.05  | 1.63    | 1.89    | 109.19      | 107.47      | 142.16       | 136.65       | 4.36| 2.26             |
| x = 0.10  | 1.63    | 1.90    | 109.53      | 106.09      | 141.52       | 135.68       | 4.72| 2.30             |
| x = 0.15  | 1.63    | 1.90    | 110.11      | 105.68      | 142.58       | 134.52       | 4.92| 2.30             |
| x = 1.00  | 1.63    |         | 111.21      | 104.86      |              |              | 5.5 | 4.07             |

Figure 1. (a) Si–O and (b) Zr–O pair distribution functions for x = 0.02 and x = 0.15. Dashed lines show the coordination of each species as a function of the distance. The decomposition of the first peak of the distribution among the contributions of the nearest neighbors are shown on the right for both (c) Si–O and (d) Zr–O and (e) Si– and (f) Zr-centered partial bond angle distributions. Solid lines are for x = 0.02, and dashed lines are for 0.15. The decomposition of the distributions among the contributions of the triangles formed by the nearest neighbors is shown on the right for both (g) O–Si–O and (h) O–Zr–O.

(B) Optimization: A forcefield adequate for the material and application is chosen, and its relevant parameters are optimized. The parameter optimization is driven such that the MD simulations using the forcefield would accurately estimate the *ab initio* results obtained in step (A). Some applications may warrant a validation of the optimized forcefield, particularly when the basis forcefield is a newly developed one.

One can then apply the resulting force field in MD simulations to discover or ascertain particular material characteristics. For this purpose, two MD simulations steps are necessary:

(C) Sample Material Development: MD is used at this step for virtual sample preparation. This sample preparation needs to give consideration to the physical context and boundary conditions of the material of interest.

(D) Material Characterization: This stage of the algorithm amounts to running MD simulations and performing computations on the simulation results to determine yet unknown characteristics of the material of interest.

The Methods section of this article presents a detailed description of the above CAMD algorithm in our work. We then present and discuss the particular dielectric constant results we generated for Zr-doped silica glasses and the underlying atomic scale properties that govern it. A Supporting Information document complements this article. The latter provides more information and details on the following aspects of our work: forcefield description and parameters, forcefield optimization procedure and its implementation code, and dielectric constant computations.

### RESULTS AND DISCUSSION

**Structural Characterization.** First, we analyze the atomic structure of (ZrO$_2$)$_x$(SiO$_2$)$_{1-x}$ glasses for various x values. In Figure 1 we show the radial distribution functions for Si–O and Zr–O pairs for the highest and lowest Zr contents. Both radial distribution functions present a sharp peak for the first neighbors without clear secondary peaks, a common characteristic of highly disordered materials. The $g_{\text{Si–O}}(r)$ is almost identical for both compositions, which indicates that the local tetrahedral arrangement of the silicate groups is not affected by the Zr content. On the other hand, there are significant changes in the $g_{\text{Zr–O}}(r)$: the first peak is shifted to larger values (see Table 1) and the average coordination number (CN) increases from $\approx$4 to $\approx$5. A more detailed analysis decomposing the first peak into partial contributions reveals that Zr is mainly 4-coordinated [Zr(IV)] in the 0.02% Zr sample and evolves to 4,5-coordinated [Zr(IV) and Zr(V)] environments with a considerable 6-coordinated [Zr(VI)] contribution. We also observed the presence of 3-fold coordinated oxygen atoms, known as "tricenter" oxygen, similar to what has been reported for silicate and aluminosilicate glasses.13–15

The angular distribution functions for the O–Si–O and O–Zr–O angles presented in Figure 1 also reflect the change in the coordination environments. The O–Si–O angles are centered at 109°, corresponding to the tetrahedral geometry, without significant differences for different Zr contents. In contrast, the O–Zr–O angle distribution is considerably wider, with a maximum value that shifts toward lower angles as the Zr content increases. That angle decrease is consistent with the evolution of some Zr(IV) sites from tetrahedral coordination to higher CN...
Zr-average CN with the concentration of Zr, consistent with the experimental observations using IR spectroscopy from Lucovsky and Rayner. As observable in Figure 2, we obtained a good linear fit (CN = a + bx) with the parameters a ≈ 4.00 and b ≈ 6.42. The extrapolation to x = 0.5 in the linear fit leads to (CN) ≈ 7.2 which is the expected average CN for Zr in amorphous zircon.

Other structural parameters are detailed in Table 1 and have been discussed earlier in the text.

**Dielectric Constant.** The main objective of this work is to compute the dielectric constants of (ZrO2)x(SiO2)1−x from the dipole moment fluctuations. Before discussing the final values, it is important to pay attention to the convergence to a stable/asymptotic value of the dielectric constant. As shown in Figure 3, these convergence rates varied for different compositions. The convergence was faster for pure SiO2 than those for the (ZrO2)x(SiO2)1−x compounds, and increasing the ZrO2 content resulted in larger convergence times. The reason for the slower convergence for these materials is the relatively complex structure of (ZrO2)x(SiO2)1−x compounds, resulting in a larger variance in the total dipole moment and therefore requiring large times for the averaging to reach asymptotic values for the dielectric.

Figure 3 shows our results for five different glasses with variable SiO2 to ZrO2 ratios. First, we evaluated the method comparing the dielectric constant ε obtained from MD with the empirically known values for the end member oxides SiO2 and ZrO2. Our result for SiO2 ε = 3.75 is in excellent agreement with the experimental range of 3.7−3.9, and for ZrO2 at ε ≈ 22.5, our value is within the experimental range 20−24. Then, we computed the dielectric constant for four arbitrary (ZrO2)x(SiO2)1−x compositions to showcase the ability to estimate their dielectric constants numerically. The dielectric constant increases with the Zr content from the value of pure SiO2 up to 15.6. An equivalent increase in the dielectric has been observed experimentally yet our simulations overestimate the absolute value. Nevertheless, the qualitative increase with the Zr content is properly captured, which postulates the presented classical MD simulations as a predictive tool for the calculation of dielectric properties of highly disordered materials.

To explain the atomic scale origin of the ε increase with the Zr content, we explored two parameters that play a role in the dipole moment fluctuations within the classical atomic description adopted here, namely, the vibrations of the atomic positions and the atomic charges. The atomic positions are related to the vibrational frequencies of the material, and lower vibrational frequencies are associated with higher static dielectric constants. To study them, we have quantified the flexibility of the glass three-dimensional framework relying on the topological constraint theory (TCT). This approach has already been successful at describing a wide variety of glass properties, including their dielectric response. TCT reduces the description of complex glass networks to a set of nodes (atoms) connected following a set of constraints defined by the chemical bonds. These constraints are divided in two sets: radial (or bond stretching, BS) constraints and angular (or bond bending, BB) constraints. Thus, the rigidity of the system is described by the average number of constraints per atom nα where nα = 3 sets the isostatic limit, nα > 3 corresponds to a stressed-rigid domain, and nα < 3 corresponds to flexible glasses.

Following the method proposed by Bauchy et al., the constraints are computed directly from the MD simulations by studying the pair distribution functions and partial bond angle distributions of each atomic species. The partial pair distribution
functions and bond angle distributions showed in Figure 1 allow to decide which radial and angular constraints are satisfied: a narrow distribution suggests an active constraint, whereas a broad one implies a broken constraint. Therefore, the standard deviation of each distribution must be computed. These are presented in Figure 4 for the radial and angular distributions. In order to do an appropriate counting, we studied independently the distributions of Zr according to their coordination, that is, Zr(IV), Zr(V), or Zr(VI), as well as O(II) and "tricluster" O(III). Since each Si or Zr atom is bonded to an oxygen atom, the oxygen atoms have been excluded from the radial constraint counting, assigning the full constraint to the cation.\(^{26-28}\)

Considering all previous data, the radial standard deviations in Figure 4 show a clear gap between the deviations of the atoms within the first coordination shell and the second. Therefore, the number of BS constraints for each kind of atom is defined by its coordination: 4 constraints for Si and 4, 5, 6 for Zr(IV), Zr(V), and Zr(VI), respectively.

Regarding the angular constraints, the limit between broken and intact angular constraints is set around \(\sigma = 15.27\) As\(^{\circ}\). Expected, O(II) has a single angular constraint, while "tricluster" O(III) has 2. Similarly, six angles around Si atoms are constrained, which correspond to the angles formed by its four neighbors. In this case, however, only five of these angles are independent,\(^{29}\) leading to five angular constraints. Finally, similar to Al in aluminosilicates, no angular constraint is assigned to Zr atoms due to the large variance of its partial angle distributions.

Once this analysis was done, we computed the average number of constraints per atom \(n_c\). It must be noted that the BS and BB constraints for each atomic species remain constant for the different compositions, and the total number of constraints per atom changes only due to the relative proportion of O(II)/O(III) and the inclusion of more Zr without BB constraints. Denoting \(z_s\) and \(z_o\) as the fraction of Zr(V) and Zr(VI) with respect to the total Zr, the average number of constraints is

\[
n_c = \frac{11}{3} + x \frac{3z_s + 6z_o - 5}{3}
\]

Figure 5 depicts the relationship between the amount of Zr and the number of constraints. The number of constraints decreases as more Zr is added, therefore becoming more flexible, which is consistent with the \(\epsilon\) increase. As mentioned, the change of \(n_c\) is due to the increase of Zr and O(III) in the structure. However, both species have a contrary effect on the flexibility: Zr, without BB constraints, induces higher flexibility, while O(III) has an extra BB constraint compared to O(II), inducing rigidity. As a result, the mentioned increase of the flexibility is not very important, with \(n_c\) decreasing from 3.64 to 3.57. For comparison, the number of constraints may range from 2.8 to 3.5 when decreasing the Ca/Si rate in C-S-H\(^3\) or from 2.9 to 3.4 for glassy aluminosilicates with Al molar fractions changing from 0 to 17%.\(^{14}\)

We now turn to evaluate the atomic charges as a function of the Zr content. The charge distribution computed by the electronegativity equalization method (EEM) method for the different atomic species is shown in Figure 6. The obtained charges are consistent with the expected values and with the phenomenological rule for the charge of mixed oxides proposed by Barr et al.\(^33\) They concluded from X-ray photoelectron spectroscopy experiments that in mixed oxides, the cation becomes more covalent than in the pure oxide and the cation becomes more ionic than in the pure oxide. Our simulation...
agrees with this empirical rule. From Figure 6, the charge distribution on Zr atoms shifts toward higher positive values at high Zr contents from $\approx 2.25$ to $\approx 2.5$, while the charge distribution of Si atoms shifts to lower positive values. The charges on oxygen atoms do not change substantially with the amount of Zr, with a narrow distribution around $\approx -0.85$. The shift toward higher and lower charges for the two cations induces a larger heterogeneity in the local charge distribution that might increase the dipole moment fluctuations and consequently the static dielectric constant.

It is noteworthy that the charge decrease on Si atoms takes place by a shift in the intensity of two well-defined peaks. In Figure 6c, we show the variation of charges on the Si atoms around the central value of 1.25 e$^-$. Clearly, the population of the peaks is directly related to the amount of Zr in the system, and it is homogeneously distributed across the simulation box. However, we could not identify any relationship with the local chemical environment around each Si. Further research would be needed to understand these effects or identify if it could be a spurious result from the EEM method.

## CONCLUSIONS

In this work, we reported a computer-aided characterization of dielectric materials of potential value for the CMOS technology. More specifically, we investigated ZrO$_2$ and (ZrO$_2$)$_x$(SiO$_2$)$_{1-x}$ glasses with variable Zr contents using classical MD with ReaxFF empirical potentials.

First, we merged two ReaxFF sets (Si/O/H and Zr/O/H) and parameterized non available cross terms with a home-made code to build a new Si/O/Zr/H set. Then, we prepared (ZrO$_2$)$_x$(SiO$_2$)$_{1-x}$ glasses and computed the dielectric constant from the average dipole moment fluctuations of the simulation box. We find results in qualitative agreement with the available experimental data: as the Zr content increases, the dielectric constant of the amorphous (ZrO$_2$)$_x$(SiO$_2$)$_{1-x}$ steadily increases. Analyzing our simulations, we can attribute such an increase to two factors. On the one hand, there is an increase of the framework flexibility as the Zr content increases, that we have quantified from the point of view of the TCT. On the other hand, there is an increasing heterogeneity of the charge in the cations, with Zr becoming more ionic and Si more covalent as the Zr content increases. Those two factors, from which the latter is the more marked, induce larger fluctuations in the dipole moment and, consequently, a larger static dielectric constant.

To the best of our knowledge, our work is the first reported use of classical MD simulations for predicting the dielectric constant of amorphous solids. Our work and findings illustrate an example of how MD-based CADM can constitute an effective and attractive methodology for advancing complex CMOS technology which is highly dependent on the intricate interaction of a large number of well-known and ill-studied materials.

## METHODS

**Ab Initio Simulations.** We used density functional theory (DFT) to perform calculations of small clusters containing Si, O, Zr, and H. The Gaussian-16 program$^{34}$ was used to perform these calculations. In order to select an accurate exchange–correlation (XC) functional to compute structures and energies, we first compared results from different XC functionals with those from the MP2 perturbation theory.$^{35,36}$ We ended up using the B3LYP functional with the D3 version of Grimme’s dispersion.$^{37}$ Equilibrium Zr–O–Si angles obtained using B3LYP were equal to those calculated using MP2, providing evidence of the accurate predictive capability of this functional for our application. Si, O, and H were described at the full
electron level by a variation of the 6-31G basis set with single diffuse functions as well as d functions. Zr was described using an energy-consistent relativistic pseudo-potential along with the correlation consistent basis set with diffuse functions. This pseudo-potential/basis set pair is specific for 4d transition metals and is available through the basis exchange database under aug-cc-pVTZ-PP.

In the end, we calculated the electronic energy for two systems subjected to variations: one for a system with Zr−O−Si angles varied across a range of values away from the equilibrium angle and a second system with hydrogen bonds also varied over a range of values away from the equilibrium bond length. More details on the molecular systems and \textit{ab initio} simulations are provided in the Supporting Information document. These two atomic structures and their corresponding energies formed the basis for the next step.

ReaxFF Parametrization. We selected ReaxFF as our choice of classical forcefield because of its broad applicability. For the specific problem of dielectric constants, ReaxFF is suitable due to the EEM which let us compute dipole moment fluctuations due not only to atomic displacements but also to charge fluctuations. ReaxFF was initially introduced to study hydrocarbon systems, yet ReaxFF parameter sets were later developed and used to successfully describe various properties of semiconductors and transition metals as well as various processes related to those systems.

The ReaxFF parameters for the Si/Zr/O/H set we parameterized are based on the independent sets made by Fogarty et al. for SiO\textsubscript{2} and van Duin et al. for ZrO\textsubscript{2}. Since electronegative elements such as Hf and Zr tend to form bonds with oxygen rather than Si, parameterizing the Zr−O−Si angular terms seems sufficient to accurately model both bulk and interface cases with (ZrO\textsubscript{3})\textsubscript{4}(SiO\textsubscript{2})\textsubscript{1−}\textsubscript{y} at least for low concentrations, when Zr acts a network modifier.

We optimized our ReaxFF parameters using the molecular system calculated by DFT. The optimization criterion was to minimize the mean-square error of the energy computed using ReaxFF. The criterion was to set the energy-consistent relativistic pseudo-potential and the \textit{ab initio} calculations for the same sample.

Preparation of \textit{(ZrO\textsubscript{2})\textsubscript{1−}x(SiO\textsubscript{2})\textsubscript{x}} Amorphous Systems. Using the optimized forcefield produced by our Pyfield tool, we performed MD simulations using LAMMPS (2018 release). The following steps were taken to prepare the system:

- Placement: The systems we studied contain 24,000 atoms, 2 orders of magnitude larger than previous attempts to compute static dielectric constants with DFT methods, which lead to a better sampling of the properties. The amorphous \textit{(ZrO\textsubscript{2})\textsubscript{1−}x(SiO\textsubscript{2})\textsubscript{x}} systems were prepared based on a technique previously used by randomly placing 8000 SiO\textsubscript{2} molecules in a box with the dimension of 72 × 72 × 72 Å\textsuperscript{3} and randomly replacing a portion of Si atoms with Zr to achieve the desired ratio. We then proceeded with annealing the system as per the following step.

- Annealing: We annealed the system using an NPT ensemble by initiating it at 3000 K and cooling it down to 300 K over a span of 1 ns. We then equilibrated the system at 300 K for another 100 ps.

For our NVT and NPT simulations, we used a Nose−Hoover thermostat and barostat. The integration time steps we used for the Velocity-Verlet algorithm were set to 0.1 fs for the above steps (Figure 7).

Material Characterization. The ultimate objective of our work was to determine the permittivity of Zr-doped silica glasses which we assumed to be homogeneous and isotropic. The standard definition of polarization \( P \) is such that \( P = \epsilon_0 \chi \cdot E \), where \( E \) is the applied electric field, \( \epsilon_0 \) is the vacuum permittivity, and \( \chi \) is the dispersion. Ideally, we would want to apply an electric field, collect the polarization data, and compute \( \chi \) which would lead to the estimation of the relative permittivity \( \epsilon = \chi + 1 \). However, applying an electric field to a classical MD simulation is not trivial. Instead, a reasonable approach is to use dipole fluctuations to calculate bulk (ZrO\textsubscript{2})\textsubscript{4}(SiO\textsubscript{2})\textsubscript{1−}x Permittivity in a fashion similar to how various liquids have been studied using the expression

\[
\epsilon = 1 + \frac{\langle (M^{(E=0)})^2 \rangle - \langle M^{(E=0)} \rangle^2}{3k_BT/V \epsilon_0}
\]

where the \( \langle (M^{(E=0)})^2 \rangle - \langle M^{(E=0)} \rangle^2 \) term denotes the mean-square deviation of the total dipole moment, \( k_B \) is Boltzmann’s constant, \( T \) is the temperature, and \( V \) is the volume over which the computation is performed. To compute the ensemble averages \( \langle (M^{(E=0)})^2 \rangle \) and \( \langle M^{(E=0)} \rangle \), we performed the simulations based on the following procedure:

- Initiation: Systems are initiated and equilibrated at 300 K for 200,000 steps where each time step is 1 fs.
- Fluctuation: Systems were allowed to fluctuate using an NVT ensemble at 300 K over a span of 8 to 12 ns depending on the sample with a time step of 1 fs.

The total dipole moment of the system was computed during the fluctuation step at every 20 ps, and we used them to compute the relative permittivity as a function of time. The dipole moments are computed according to \( M = \sum q_i \vec{r}_i \) where \( q_i \) is the partial charge of a particle and \( \vec{r}_i \) is the location vector of the charge from an arbitrary origin.

## APPENDIX

In this section, we provide a simple derivation of eq 2. The derivation should help the readers understand the limits of its applicability. Assuming an electric field in the \( \vec{x} \) direction

\[
\vec{P}_x = \frac{\langle (M^{E=0})^2 \rangle - \langle M^{(E=0)} \rangle^2}{V}
\]
where \( \langle \rangle \) stands for ensemble average, \( M_z \) is the dipole moment in the \( x \) direction, and \( V \) is the system’s volume. Using the Boltzmann law, we have

\[
\langle M_x^E = E_f - M_x^{E_0} \rangle = \int \cdots \int d_{\text{p}_1} \cdots d_{\text{p}_N} (M_x^{E = E_f} - M_x^{E_0}) \exp\left( -\left( U - M_x^{E = E_f} \right)/kT \right) / \int \cdots \int d_{\text{p}_1} \cdots d_{\text{p}_N} \exp\left( -\left( U - M_x^{E_0} \right)/kT \right)
\]

(4)

where the integration is over all degrees of freedom of all of the particles and \( U \) is the energy of the system when no electric field is applied. Performing a Taylor series expansion of eq 4 at \( M_x^{E_0} \) and only keeping the first-order term yields

\[
\langle M_x^E = E_f - M_x^{E_0} \rangle = \frac{\partial}{\partial E_f} \left( \int \cdots \int d_{\text{p}_1} \cdots d_{\text{p}_N} (M_x^{E = E_f} - M_x^{E_0}) \exp\left( -\left( U - M_x^{E = E_f} \right)/kT \right) \right) / \left( \int \cdots \int d_{\text{p}_1} \cdots d_{\text{p}_N} \exp\left( -\left( U - M_x^{E_0} \right)/kT \right) \right) \times E_x \times E_x
\]

(5)

where the last equality is valid if the material is isotropic.\(^a\) By substituting the right-hand side of eq 5 with \( \langle M_x^E = E_f - M_x^{E_0} \rangle \) in eq 3 and considering that \( \bar{P}_x = \epsilon \partial \bar{E}_x \), we have

\[
\chi = \frac{\langle M_x^E = E_f - M_x^{E_0} \rangle}{3kTv_0}
\]

(6)

and \( \epsilon = \chi + 1 \).

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