

Molecular Dynamics Simulations of Gas Selectivity in Amorphous Porous Molecular Solids

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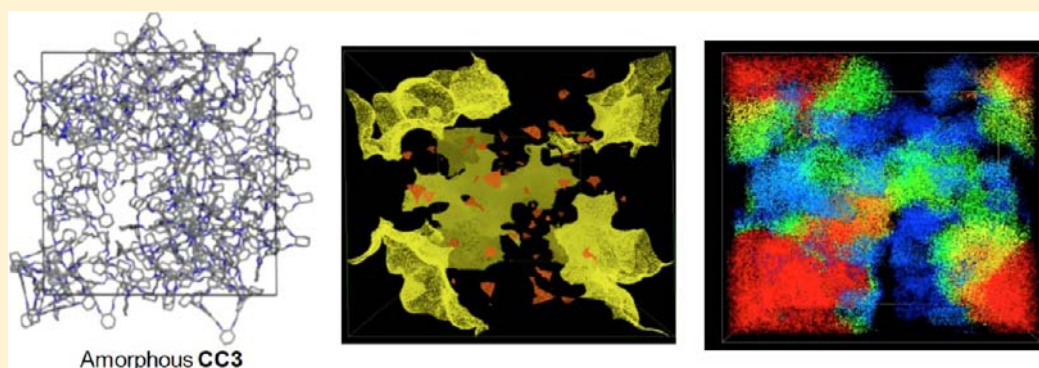
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S Supporting Information



ABSTRACT: Some organic cage molecules have structures with protected, internal pore volume that cannot be in-filled, irrespective of the solid-state packing mode: that is, they are intrinsically porous. Amorphous packings can give higher pore volumes than crystalline packings for these materials, but the precise nature of this additional porosity is hard to understand for disordered solids that cannot be characterized by X-ray diffraction. We describe here a computational methodology for generating structural models of amorphous porous organic cages that are consistent with experimental data. Molecular dynamics simulations rationalize the observed gas selectivity in these amorphous solids and lead to insights regarding self-diffusivities, gas diffusion trajectories, and gas hopping mechanisms. These methods might be suitable for the de novo design of new amorphous porous solids for specific applications, where “rigid host” approximations are not applicable.

1. INTRODUCTION

Molecular selectivity is central to many applications of nanoporous solids,^{1–3} and this is determined by the size and shape of the pores and interactions with the pore surfaces.⁴ Control over pore structure is therefore desirable, so that predictive structure–property relationships can be established. In recent years, crystalline porous solids, such as zeolites and metal–organic frameworks (MOFs),³ have played a dominant role here. These ordered structures have uniform pore sizes that can be characterized at the molecular level using techniques such as X-ray diffraction. Hence, a range of computational techniques have been developed to provide a rational, molecular-level design basis for crystalline porous solids for separations.^{5–9} By contrast, there are fewer computational approaches to underpin the molecular design of amorphous porous solids, despite their possible practical advantages. For example, amorphous porous polymers can form robust, solution-processable separation membranes.¹⁰

Recently, porous molecular solids,^{11–13} where packing is dictated by weak van der Waals forces rather than extended bonding, has attracted attention. Unlike extended networks, “porous molecules” can be soluble in organic solvents and have been solution processed into nanoparticles¹⁴ as well as porous composites¹⁵ and molecular sensors.¹⁶ So far, most porous molecular solids, other than nanoporous polymers,¹⁰ have been crystalline. However, amorphous porous molecular solids^{17–20} can have practical advantages. For example, some organic cage molecules are much more porous when rendered amorphous than they are in their crystalline forms.¹⁴ We rationalized this on the basis of additional, extrinsic porosity between the cage molecules that is not present in the crystalline state. It is possible to process organic cage molecules into amorphous solids from solution: for example, by freeze-drying.¹⁴

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Amorphous porous molecular solids can also be created by chemical synthesis, for example, by synthesizing “scrambled” organic cages comprising a mixture of molecules with different shapes that cannot therefore crystallize, irrespective of how the solvent is removed.^{14,20} However, the purposeful molecular design of such materials is challenging because, unlike crystalline porous solids, functions such as sorption selectivity cannot simply be correlated with a crystal structure. Hence, there is a need to develop underpinning computational methods for the molecular level understanding of porosity in amorphous molecular solids.²¹

A primary challenge in simulations for amorphous solids is to generate physically representative models. This is more difficult than for ordered materials where crystal structures are known. For amorphous polymers, a common approach is to pack polymer chains at either an artificially low density or to an observed experimental density and subsequently to follow several compression and relaxation steps using molecular dynamics (MD) simulations until the target experimental density is achieved.^{22–26} However, some drawbacks exist for these methods. Most significantly, the simulations rely on an experimental target density. This can be challenging to determine accurately for microporous solids, and the density is, of course, unknown for new, hypothetical molecular designs. In some cases, matching to an experimental target density might lead to oversimplifications because inhomogeneities in the solid, for example, large voids, cannot be represented in a small simulation cell while still matching the bulk experimental density. Hence, poor estimation of physical properties might arise.^{24,27} In addition, these procedures sometimes fail to build a realistic structure due to atomic overlaps of large, rigid repeat units.²⁵ Nonetheless, a number of computational studies exist for amorphous porous organic polymers, such as polymers of intrinsic microporosity (PIMs),¹⁰ hypercrosslinked polymers (HCPs),^{28–30} and conjugated microporous polymers.³¹ Of particular note, Colina and co-workers reported a simulation scheme for generating models for HCPs that followed the synthetic polycondensation route for these materials.^{30,32} The final step in structure generation was followed by a 21-step compression and slow decompression protocol that resulted in densities that were close to those from experiments. Significantly, physical properties such as density and pore volume emerged from the simulation, rather than being input from experiment. The same approach was used to generate structural models for amorphous PIM materials^{33,34} and structurally related amorphous porous organic molecules.³⁵ In this latter work, the molecules were packed at either a specified final density or at a low density, and a MD compression/relaxation scheme or annealing procedure was applied.³⁵ Structural analysis was carried out by comparison of simulated and experimental densities, surface areas, pore volumes and wide-angle X-ray scattering (WAXS). It was observed that different simulation methods had a significant effect on the structural models. Packing molecules at a low initial density and then performing a 21-step MD simulation was proposed as the best simulation methodology for these systems, since it was again not necessary to predetermine a target density. These simulations both rationalize known materials and suggest approaches to design new, improved molecules: for example, suggesting that more rigid structures and bulkier functional end groups might enhance microporosity. A follow-up study by Siperstein and co-workers involves the calculation of surface areas, free volumes, and argon adsorption isotherms.³⁶

A large number of computational studies have focused on gas diffusivities in amorphous polymers,^{22,27,37,38} not least because of the relevance of such phenomena in gas membrane separations. A hopping mechanism has been suggested for gas diffusion, whereby gas molecules jump from one cavity to another as a result of transient channels appearing between these cavities during dynamic motions.^{39,40} A relatively small number of simulation studies have focused on gas diffusion in porous molecular solids, as opposed to polymers, and these relate to crystalline rather than amorphous organic solids. For example, studies have investigated guest inclusion and diffusion in crystalline cucurbit[*n*]uril and calixarenes.^{41–44} In terms of crystalline organic cages, we have used MD simulations to rationalize H₂/N₂ gas selectivity for a crystalline porous cage, CC1 β ,⁴⁵ and the molecular selectivity of crystalline CC3-R toward various C8 and C9 aromatic isomers.⁴⁶ To date, however, there are no studies to our knowledge on MD of guest diffusion in porous amorphous molecular solids.

This work focuses on two isostructural organic cages, CC1 and CC3. Their molecular structures are shown in Figure 1a.

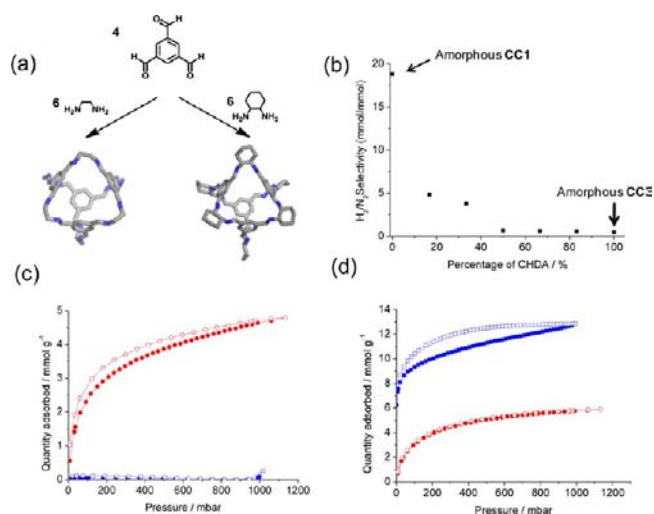


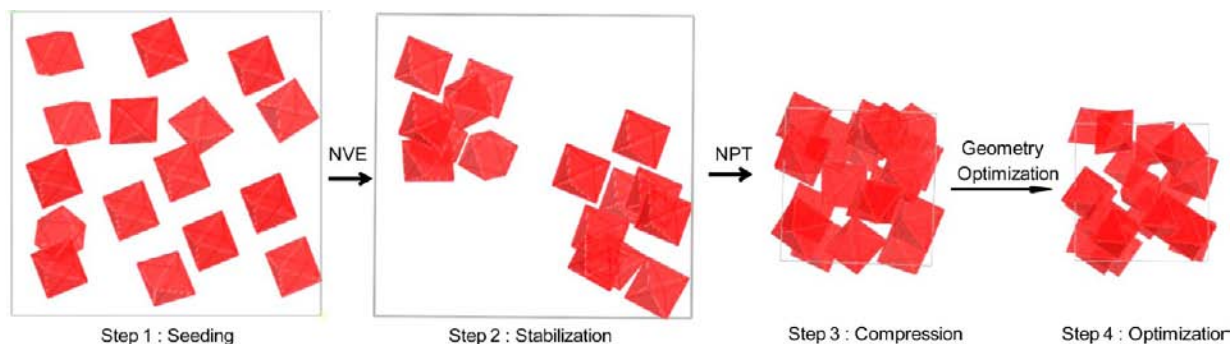
Figure 1. (a) Molecular models of CC1 and CC3 (hydrogens are omitted and carbon and nitrogen are colored gray and blue). (b) Ideal H₂/N₂ gas selectivity as a function of the ethanediamine-to-cyclohexanediamine (EDA/CHDA) ratio for “scrambled” amorphous cage materials that were synthesized with different ratios of these two diamines²⁰ and for unscrambled, freeze-dried amorphous CC1 and amorphous CC3. H₂ (red) and N₂ (blue) adsorption/desorption isotherms for (c) amorphous CC1 at 77 K and for (d) amorphous CC3 at 77 K, both prepared by freeze-drying.

Both cage molecules have tetrahedral symmetry and are synthesized via [4 + 6] imine condensation reactions of triformyl benzene and vicinal diamines: namely, ethanediamine for CC1 and cyclohexanediamine for CC3. Hence, CC1 has six “unfunctionalized” ethane vertices, while CC3 has bulkier cyclohexyl groups on its vertices.⁴⁷

2. COMPUTATIONAL AND SYNTHETIC METHODS

2.1. Challenges in Simulating Amorphous Cage Packings. In crystalline solids, it is common for the most stable molecular packing to be the densest possible, as exemplified by crystalline CC3-R, both experimentally and in crystal structure prediction studies.⁴⁸ However, for amorphous solids, the molecular packing arrangement is often metastable with respect to a denser, crystalline phase. Hence, a problem arises in that simulations must be effective in reproducing a

Scheme 1. Four-Step Simulation Procedure for Generation of Amorphous Cage Structural Models



metastable experimental state, as defined by the stability time scale for the real amorphous solid, but the density of molecular packing in the simulation must not be “artificially” low as a result of insufficient equilibration. Moreover, amorphous solids may be more inhomogeneous than crystalline phases, and local “defects”, such as voids, may exist where local density is significantly lower than the bulk average. As a result, large simulation cells and a number of different structural models over which properties can be averaged is the most effective way to simulate amorphous solids. In addition, a sufficiently long simulation time is required to allow structural relaxation and, hence, to avoid trapping of unrealistically high energy states. For all of these reasons, simulation of amorphous materials is computationally challenging.

2.2. Generation of Structural Models. A Cage Specific Force Field (CSFF), developed previously for porous organic imine cages, was used.⁴⁹ The partial charges were assigned by the force field (see Table S1, Supporting Information). All MD simulations were carried out with DL_POLY2.20⁵⁰ at 300 K, a pressure of 1 atm, the Verlet leapfrog algorithm⁵¹ and a time step of 0.5 fs. A summary of simulation parameters is shown in the Supporting Information (section 1.3, Tables S2 and S3). The amorphous models were generated using the following steps: (1) seeding; (2) stabilization; (3) compression of simulation cell; and (4) geometry optimization (Scheme 1). This approach differs from the generation of amorphous polymer structures because a single, predefined molecular unit exists for these cages, whereas a distribution of molecular chains and structures must be generated to simulate amorphous polymers. Other simulation methods for amorphous molecular solids^{35,36} contain additional compression, relaxation, and annealing steps with MD simulations, in order to sample molecular conformations and molecular packing arrangements. However, the cages are more isotropic in shape and do not exhibit such a large range of conformations, particularly in the case of CC3. They are likely to have lower rearrangement energies for molecules relative to each other because their shapes are more spherical. High-temperature annealing MD simulations were therefore not performed. In the case of CC1, there are also two different conformers with C_3 and T_d symmetry (see Figure S1, Supporting Information). Here, it was assumed that the cage exists as its more stable tetrahedral conformer in the amorphous solid, based on the fact that the C_3 conformer has only been observed in certain crystalline solvates of CC1^{45,52} and not in any desolvated forms of CC1. For reference, the energy barrier for interconversion between T_d and C_3 conformers was found to be ~ 32 kJ mol⁻¹ by both an NMR and DFT study. Both CC1 and CC3 have helical chirality; in this study, the CC1-*R* and CC3-*R* enantiomers were used for the simulations for simplicity. In reality, CC3-*R* will be a single, noninterconvertible enantiomer, while CC1 is likely to exist as a racemate. To explain the four simulation steps in more detail:

- (1) The single cage molecular structures were taken from single-crystal X-ray diffraction structures and loaded to a low density of $0.15\text{--}0.2$ g cm⁻³ using the Universal Force Field (UFF)⁵³ in the Amorphous Cell module of Materials Studio Modeling 5.0 (Accelrys).⁵⁴ The 40 cage molecules were treated as rigid bodies. Six different initial configurations were generated for

each system (see Tables S4 and S5, Supporting Information). Larger cell sizes, with 50 and 60 cages, were also generated (section 1.5, Table S4, Figure S2, Supporting Information) to confirm that calculated structural properties were not affected by the artificial amorphous cell size of 40 cages.

- (2) An MD simulation using a microcanonical NVE ensemble for 500 ps was used to stabilize the low density structure from step 1. The simulation was performed with a Nosé-Hoover thermostat and barostat.^{55,56} The cage molecules were fully flexible and clustered so as to maximize favorable intermolecular interactions.
- (3) An NPT MD simulation was run with the Berendsen thermostat and barostat⁵⁷ for 8 ns. The Berendsen thermostat offers a good scheme for controlling temperature when the initial system is far from the equilibrium.^{58,59} The volume and configuration energy fall substantially at the beginning of the run and then maintain a constant average after ~ 8 ns (see Figure S3, Supporting Information). A further 7 ns simulation was carried out for one CC1 and CC3 model to confirm that no further volume compression was observed (see Figure S4, Supporting Information).
- (4) The last configuration from step 3 was geometry optimized using the conjugate gradient method in the Discover module in Materials Studio. The final cell lengths were 38.6 ± 0.5 Å for CC1 and 45.0 ± 0.9 Å for CC3 (see Tables S5–7, Supporting Information).

2.3. Simulation of Gas Diffusion. MD simulations were performed to study N₂ and H₂ diffusion in fully flexible amorphous CC1 and CC3. One gas molecule was placed in each structural model for diffusion analysis. The simulations were run using the canonical NVT ensemble. The simulation cells were equilibrated for 50 ps with the Nosé-Hoover thermostat⁵⁶ at 300 K and a time step of 0.5 fs. Both of the gas molecules were described as linear rigid molecules (see further details in section 1.6, Tables S8–10, Supporting Information). The potential parameters for H₂ were obtained from the CSFF without charge consideration. The bond length of H₂ was 0.74 Å, and a three-site linear multipole with a bond length of 1.09 Å was used to describe the N₂ molecule. The N₂ center of mass had a partial charge of $q = +0.964$ and the nitrogen atoms a charge of $q = -0.482$. The N₂ nonbonding interactions were taken from Potoff et al.⁶⁰

To calculate the self-diffusivity, the mean square displacement (MSD) was calculated and then the Einstein equation (eq 1) allowed the self-diffusion coefficients for N₂ and H₂ to be calculated during a 20 ns NVT simulation.

$$D_s(c) = \lim_{t \rightarrow \infty} \frac{1}{6t} \langle |\vec{r}(t) - \vec{r}(0)|^2 \rangle \quad (1)$$

Here, $\vec{r}(t)$ is the position vector for the diffusing molecule at time t , and $\vec{r}(t) - \vec{r}(0)$ is the vector distance traveled by a diffusing molecule over a time interval of the length t . Normal diffusion occurs when the slope of the logarithmic plot is close to 1.0, and these linear

regions were used to calculate the self-diffusion coefficients for all models (section 1.7, Table S11, Supporting Information).

The trajectory of the gases was analyzed over 10 ns of the NVT simulation to investigate the diffusion mechanism. The location of the N_2 and H_2 molecules during the MD simulations was identified by calculating the distance between the center of mass of the gas and each cage molecule. If the gas was less than 3.5 Å from the closest cage center of mass, then it was defined as being inside that cage, while for values between 3.5–4.5 Å, it was defined as being in the cage window. If the gas was not inside any cage or window, then it was defined as being in the extrinsic pore volume (Figure S5, Supporting Information). This also allowed us to calculate the proportion of the cages that were occupied over the 10 ns simulation.

2.4. Analysis of Pore Connectivity and Interconnected Surface Area (ISA). Comparison of the calculated geometrical surface area and pore volume for models with experimental values derived from gas sorption measurements is a useful, albeit imperfect, way to validate structural models for amorphous porous solids. In general, two types of calculated surface area are commonly used: accessible surface area (ASA) and Connolly surface area (CSA). CSA is defined as the interface between the surface of a probe and the pore surface, while ASA is the surface created by the center of a probe (see Figure S6, Supporting Information).⁶¹ The ASA can be used to calculate the surfaces of cavities where a probe fits, but without considering pore connectivity and whether a probe could percolate through a cell to that void.⁶¹

Unconnected “pockets” of pore volume, into which a probe molecule is too large to travel, are unlikely to be expressed as real, physical porosity in solids. We therefore chose to calculate a quantity that we refer to as the interconnected surface area (ISA), as calculated using Zeo++.⁶² Zeo++ uses a Voronoi decomposition to obtain a representation of the topology and geometry of the void space in a porous material. Then, for a given probe size, it segments the void space into interconnected and unconnected regions. The first corresponds to a pore system that allows diffusion of the probe from one cell to another, while the second corresponds to pores that limit the probe to local regions. Zeo++ can then determine and visualize the proportion of the total surface area that is interconnected—the ISA—and the remainder, which is unconnected. Zeo++ also allows us to calculate the pore limiting diameter (the largest probe that can percolate across a cell), the largest spherical void in a structure, and the pore size distribution (PSD).

It should be noted that these definitions of “interconnected” and “unconnected” pores were originally introduced to characterize rigid, crystalline structures. They are adopted here to characterize snapshots of time-evolving amorphous structures, and one can expect that certain pores will change their character over the time of the simulation; for example, “unconnected” pores becoming transiently “connected” as a result of molecular dynamics in the solid. Although the current version of Zeo++ does not allow precise tracking of the dynamic connectivity of pores over the simulation time, we can use discrete snapshots of the pore connectivity to give a simplified view of this complex dynamic process. For example, we interpret interconnected pores (and positive values of the corresponding interconnected surface area and pore volume) as an indication of pore systems extending over the distances much larger than the studied cell. Therefore, structures having a larger contribution of ISA (and/or the corresponding interconnected volume) over the time of the simulation are expected to have higher diffusion rates because guest diffusion will not have to depend on the dynamic opening of windows in the pore structure. Likewise, a simulation with a low or zero ISA, but a significant unconnected surface area, means either that the diffusion is not possible or is expected to be slow and dependent on dynamic pore opening events.

2.5. Experimental Methods. To generate amorphous solids, samples of CC3-R and CC1 were dissolved fully in dichloromethane (5 mg/mL) before being frozen rapidly in liquid nitrogen. The frozen dichloromethane was then removed by freeze-drying, using either a commercial freeze drier (Heto, Lyolab3000) in the case of CC3-R or by high vacuum on a Schlenk line, with a liquid nitrogen cold trap, in the case of CC1. The aim was to induce rapid precipitation from

solution, followed by removal of the frozen solvent, in order to prohibit any mobility of the cage molecules required for crystallization and thus to render the material in an amorphous state. The amorphous character of the products was confirmed by powder diffraction (Figure S7 and S8, Supporting Information). Gas sorption and powder X-ray diffraction analyses were carried out as described previously.¹⁴ Given the difficulty in validating structural models for amorphous solids from gas sorption data alone, X-ray structure factors were also analyzed and compared with these models. This is similar to the approach previously used by Colina et al. to validate their amorphous molecular packing methods.³⁵ The pair distribution function (PDF) and total structure function $S(Q)$ for each structural model for amorphous CC1 and CC3 were simulated using ISAACS.⁶³ An average simulated structure function for the six CC3 models was then generated, and these average data were used for comparison with the experimental data.

3. RESULTS AND DISCUSSION

3.1. Experimental Gas Selectivity. In our previous study for scrambled cages (see Figure S9, Supporting Information),²⁰ a material with a 5:1 ratio of EDA/CHDA had the highest ideal H_2/N_2 gas sorption selectivity of around 5. Here we show (Figure 1b,c) that unscrambled, amorphous CC1 (6 × EDA vertices), as prepared by freeze-drying, is in fact much more selective and has an ideal H_2/N_2 selectivity of 19 at 77 K (1 bar). By contrast, amorphous CC3 shows little selectivity between these two gases (Figure 1d). Given this strong difference in gas selectivity between amorphous CC1 and amorphous CC3, we decided to rationalize this at the molecular level via simulations. Moreover, given the relatively ‘soft’ nature of these materials, rigid models, which assume fixed atomic positions, might be inappropriate. This is a particular issue in microporous materials where pore sizes are comparable with the size of the gaseous guests.

3.2. Structural Models for Amorphous CC1 and CC3. In order to generate a range of representative amorphous structures, and to allow averaging of physical properties between models, six independent models were constructed both for amorphous CC1 and for amorphous CC3, following the procedures described above. A comparison of the experimental crystalline structures for CC1 and CC3, and a representative amorphous simulated structure for each of these cages, is given in Figure 2.

The six independent models are labeled as AC1–M1 to AC1–M6 and AC3–M1 to AC3–M6 for amorphous CC1 and CC3, respectively. To ensure that the models obtained at the end of step 4 were representative of the experimental bulk materials, each model was characterized in terms of structure factors, surface area, and pore volume. Likewise, six different synthetic samples of amorphous CC3 (AC3–S1 to AC3–S6) were prepared for N_2 adsorption analysis (Figure S10, Supporting Information) to evaluate the inherent variation in porosity for samples prepared by the freeze-drying procedure.

X-ray Structure Factors. The X-ray structure factors from experimental samples were compared with those simulated from our models. The pair distribution function (PDF) and total structure function $S(Q)$ for each structural model for amorphous CC1 and CC3 were simulated using ISAACS⁶³ (see Figures S11 and S12, Supporting Information). An average simulated structure function for the six CC3 models was then generated, and these average data were used for comparison with the experimental data, as shown in Figure 3. A larger scale plot showing the low Q region more clearly is included in Figure S13 (Supporting Information). The positions of the peaks in the total structure function generated from the

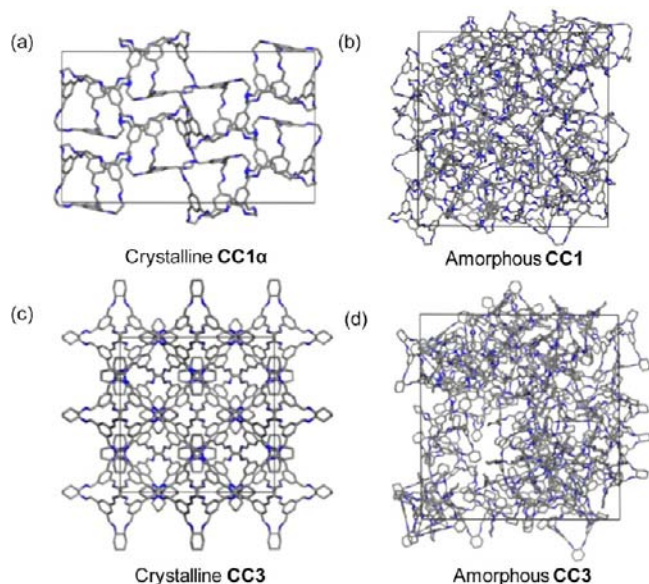


Figure 2. Structural comparison between (a) crystalline and (b) amorphous CC1 and (c) crystalline and (d) amorphous CC3. A $2 \times 1 \times 1$ super cell is shown for the crystalline CC1 α polymorph.

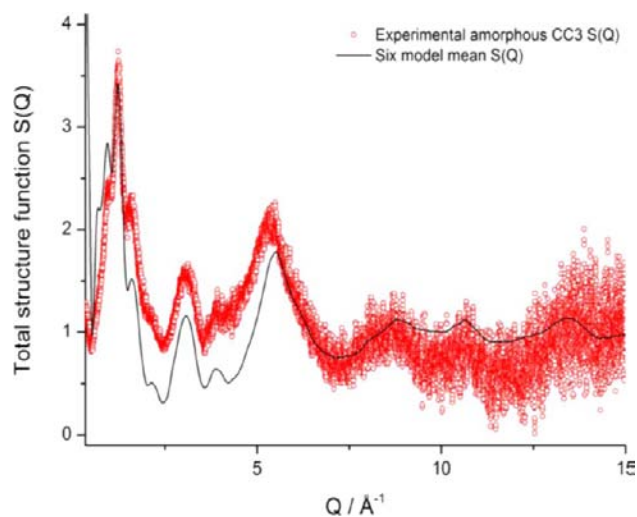


Figure 3. Total scattering for amorphous CC3 for the experimental sample (red circles) compared with simulated data that is averaged over different CC3 models (black line).

structural models are in good agreement with experimental data from bulk amorphous CC3. Furthermore, relative peak intensities and peak shapes are also consistent between the experimental and simulated scattering. For comparison, Figure S14 (Supporting Information) shows the experimental structure factor for crystalline CC3, i.e., a phase which has the same molecular structure of the cage but a different packing arrangement.

Good agreement was also observed for the simulated and experimental PDF derived from the structure function for amorphous CC3, as shown in Figure S15, Supporting Information. In addition, structural models generated from steps 3 and 4 did not show a significant effect on the structure factors, as shown in Figure S16, Supporting Information. These results suggest that the structural models for CC3 are representative of the real experimental solids. It should be noted that the same X-ray analysis for CC1 was not conducted

because of a tendency for small amounts of CC1 to crystallize from the amorphous samples over time. The amount of crystalline material is small, at least over short time scales of storage, and not sufficient to impact the bulk gas sorption measurements to any significant degree. The minor crystalline phase in CC1, however, scatters strongly, which affects the X-ray analysis. The fact that amorphous CC3 is much more stable against crystallization, even over prolonged storage times, is significant and likely reflects the less spherical nature of this molecule and its propensity to interlock in the solid state. We suggest that this translates into a much larger energy barrier for solid-state reorganization into an ordered, crystalline phase, and hence CC3 is, by virtue of its shape, much more stable as an amorphous solid than CC1.

Comparison with Experimental Gas Sorption Data. Experimental surface area and pore volume were calculated from the N₂ adsorption isotherms using the Brunauer–Emmett–Teller (BET) method for surface area and the *t*-plot method for pore volumes. It has been noted by others that experimental surface areas and pore volumes can vary widely depending on the pressure range chosen for analysis.⁶¹ Likewise, values of calculated surface area and pore volumes depend on the diameter of a probe, and the van der Waals radii of atoms of the host molecules used. Hence, an exact agreement between theory and experiment is unlikely and would in any case have limited significance, but it is generally accepted that a comparison of these quantities is a valid method for evaluating structural models for porous solids.^{33,61}

The calculated densities, surface areas, and pore volumes of the six independent simulation models obtained from step 4 for amorphous CC3 are given in Table 1. The equivalent

Table 1. Structural Properties Calculated for Six Independent Simulated Models for Amorphous CC3 (after Step 4)

	density (g cm ⁻³)	ISA (m ² g ⁻¹)	total SA (m ² g ⁻¹)	total pore volume (cm ³ g ⁻¹)
AC3–M1	0.79	992	1121	0.123
AC3–M2	0.80	931	1007	0.116
AC3–M3	0.85	0	790	0.074
AC3–M4	0.78	955	1058	0.115
AC3–M5	0.89	0	588	0.030
AC3–M6	0.76	1029	1095	0.181
avg ± σ	0.82 ± 0.05	651 ± 505	943 ± 210	0.107 ± 0.051

experimental BET surface areas, Langmuir surface areas, and *t*-plot micropore volumes are given in Table 2. The average simulated density over the six models is 0.82 ± 0.05 g cm⁻³. Total surface area (the sum of ISA and unconnected SA), the ISA, and the pore volume were generated using a N₂ probe radius of 1.82 Å⁶⁴ for each of the final configurations. The average total SA calculated for amorphous CC3 was 943 ± 210 m² g⁻¹, which is in reasonable agreement with the average apparent BET surface area of 860 ± 47 m² g⁻¹ over six experimental samples. The average of these models immediately rationalizes our previous experimental observation that amorphous CC3 has more than twice the surface area of highly crystalline CC3 (409 ± 8 m² g⁻¹).¹⁴ In the static models AC3–

Table 2. Experimental Data Obtained for Six Independent Samples for Amorphous CC3

	BET surface area (m ² g ⁻¹)	Langmuir surface area (m ² g ⁻¹)	t-plot micropore volume (cm ³ g ⁻¹)
AC3–S1	893	1073	0.32
AC3–S2	882	1133	0.28
AC3–S3	815	1025	0.27
AC3–S4	927	1156	0.32
AC3–S5	829	1040	0.28
AC3–S6	814	1020	0.28
avg ± σ	860 ± 47	1074 ± 58	0.29 ± 0.02

M3 and AC3–M5, the surface areas were not interconnected with respect to a N₂ probe radius of 1.82 Å, because the narrowest diameter of the channels (3.3 Å in both structures) was smaller than the N₂ probe, as shown in Table S12 (Supporting Information), even though large cavities also exist. For the models with an interconnected pore structure that was accessible to the N₂ probe, typically 91% of the total surface area was found to be interconnected.

Some studies have suggested that values for calculated pore volumes can vary widely when a different simulation method or probe radius is used.³⁰ Hence, we include in Table S13 (Supporting Information) a short study on the calculation of Connolly and accessible pore volumes. The radius of the probe has an effect on the calculated pore volume (Table S14, Supporting Information). The model averaged simulated pore volume was 0.107 ± 0.051 cm³ g⁻¹, as obtained from Zeo++ and Material Studio 5.0 (Table S15, Supporting Information), and a Connolly pore volume of 0.43 ± 0.07 cm³ g⁻¹ was obtained from Material Studio 5.0 (Table S13, Supporting Information). These values bracket the experimental micropore volume calculated from six separate gas sorption analyses (0.29 ± 0.02 cm³ g⁻¹).

The standard deviation in the simulated total SA, the ISA, and the simulated pore volume over the six models is higher than for the six experimental samples. This is easily rationalized because there is less inherent averaging in our simulations, even across six models, than in measurements for bulk materials. This highlights the danger in using only one model to represent the bulk and also that the predictive accuracy of these methods is constrained by computational expense.

The same analysis of surface area and pore volume for the CC1 models is shown in Table 3. Since amorphous CC1 is nonporous to N₂ at 77 K, the comparable set of experimental BET surface areas and micropore volumes are not available. The average calculated density of amorphous CC1 samples was 0.93 ± 0.01 g cm⁻³. The average total SA was 537 ± 56 m² g⁻¹, and none of this was interconnected: that is, unlike CC3, the ISA based on a N₂ probe radius was zero for all models for CC1. It should be noted that for one of the structural models (AC1–M2), the final configuration energy per cage was 2.6 kcal mol⁻¹ higher than the average configurational energy of the remaining models of 247.0 ± 0.3 kcal mol⁻¹ (see Table S9, Supporting Information). This suggests that a high energy configuration was trapped during the compression step of the structure generation procedure for AC1–M2. This model also displays a relatively high calculated pore volume of 0.054 cm³

Table 3. Structural Properties Calculated for Six Independent Simulated Models for Amorphous CC1 (after Step 4). AC1–M2 Is Not Included in the Averages

	density (g cm ⁻³)	ISA (m ² g ⁻¹)	total SA (m ² g ⁻¹)	pore volume (cm ³ g ⁻¹)
AC1–M1	0.94	0	477	0.021
AC1–M2	0.87	0	823	0.054
AC1–M3	0.91	0	595	0.022
AC1–M4	0.92	0	566	0.022
AC1–M5	0.94	0	476	0.022
AC1–M6	0.93	0	571	0.021
avg ± σ	0.93 ± 0.01	0	537 ± 56	0.022 ± 0.0005

g⁻¹, which is an outlier with respect to values for the other five models. AC1–M2 is therefore likely to be poorly representative of the amorphous CC1 structure, and it was not included in further analysis of properties. The average density over the five representative models for amorphous CC1 is 0.93 g cm⁻³, as compared to the crystallographic density for a crystalline polymorph of CC1, CC1α, of 1.033 g cm⁻³.⁴⁵ The average density over six models of amorphous CC3 is significantly lower: 0.82 g cm⁻³, in comparison to 0.97 g cm⁻³ for the crystalline CC3 structure.⁴⁷

The surface areas of interconnected channels (ISA) and each isolated pockets (unconnected SA) in amorphous CC1 and CC3 structural models are further discussed in the Supporting Information (section 7, Figure S17 and Tables S16 and S17). We carried out the surface area calculation for the MD structural snapshots to investigate the dynamic effect of ISA and unconnected SA (Table S17 and Figure S18, Supporting Information).

The simulated PSD of the six independent models for amorphous CC1 and CC3, and the experimental PSD of six different samples of amorphous CC3 are shown in Figure S19 (Supporting Information). There is a reasonable match between experimental and simulated PSDs for amorphous CC3 (Figure S20, Supporting Information).

The comparison of our models with experimental samples in terms of X-ray structure factors, surface area, density, pore volume, and PSD demonstrates that we have validated a procedure for constructing physically representative amorphous models for these molecular cage systems.

Analysis of Amorphous Molecular Packing in CC1 and CC3. The average distance between the center of mass for neighboring cage molecules is 18.8 Å for the five representative amorphous CC1 models, and 22.0 Å for the six amorphous CC3 models. To investigate the effect of cage vertex functionality on the molecular packing, we calculated the contribution of the extrinsic (between cages) and the intrinsic porosity (within the cages) to the total porosity for representative amorphous models (AC1–M4 and AC3–M4). The calculation details are described in the Supporting Information (section 1.10). In AC1–M4, 68% of the total Connolly free volume arises from extrinsic voids and 32% from intrinsic cage voids. As the surface area analysis shows, none of this is interconnected (ISA). By contrast, in AC3–M4, 86% of the total Connolly free volume corresponds to extrinsic volume and 14% to intrinsic volume in the cages. As mentioned above,

	window-to-window	window-to-vertices	window-to-arene	arene-to-arene	vertices-to-vertices
AC1-M4	17%	47%	33%	2%	0%
AC3-M4	45%	22%	25%	4%	4%

Figure 4. Molecular packing motifs for dimer pairs of cages in AC1–M4 and AC3–M4. The vertices of cage molecules are highlighted in red. AC1 has ethylene linkers on the vertices, while the vertices of AC3 are cyclohexyl groups.

90% of this surface area is interconnected. Hence, the bulky cyclohexane groups in CC3 direct the material to pack more inefficiently, creating extrinsic voids that constitute, in fact, the majority of the pore volume in this amorphous material. Furthermore, the interconnectivity of the surface area is enhanced by these extra voids. It should be noted that this tendency to create extrinsic pore volume is an inherent feature of the molecular shape of CC3, since it also expressed in its crystalline form as intercage cavities which can be occupied, for example, by guests such as iodine and osmium tetroxide.⁶⁵

The contribution of molecular shape to pore volume in these materials can be further dissected by analysis of the molecular packing motifs in models AC1–M4 and AC3–M4, as shown in Figure 4. For each model, 45 pairs of cage molecules that were “close packed”, with a cage-to-cage distance <25 Å, were analyzed for their packing mode. This showed that the dominant packing features include window-to-window, window-to-arene, window-to-vertex, arene-to-arene (π -stacked), and vertex-to-vertex motifs, as shown in Figure 4. In AC1–M4, 17% of cage–cage packing was attributed as window-to-window, 33% as window-to-arene, and 47% as window-to-vertex. The high number of window-to-arene and window-to-vertex arrangements results in pore blocking in amorphous CC1. However, in AC3-M4, window-to-window packing accounts for 45% of cage pairs, whereas the window-to-arene and window-to-vertex stacks are 25% and 22%, respectively. Arene-to-arene and vertex-to-vertex stacks are uncommon in both models, tallying with our observation for crystalline polymorphs of these cages that arene–arene π – π stacking is rarely observed.

The greater number of window-to-window motifs in the amorphous CC3 model gives a more interconnected pore volume. In essence, we believe that CC3 is structurally predisposed to form window-to-window interactions, both in crystalline¹⁴ and amorphous phases, based on interlocking of cyclohexane vertices in a manner that is analogous to the “sextuple aryl embrace”,⁶⁶ which involves interlocking aryl rings. Our earlier calculations on binding energies of pairs of cages in different packing modes showed that there was a significant energetic preference for window-to-window motifs by over 80 kJ mol⁻¹ (binding energy of -150 kJ mol⁻¹ compared to ~ -60 kJ mol⁻¹ for other motifs).¹⁴ CC1 lacks the structurally directing cyclohexane vertices, and hence lacks this propensity. This structure directing property in the CC3 cage molecule is thus both a cause of additional extrinsic porosity in the amorphous state and a means of ensuring that the porosity in the solid is highly interconnected via window-to-window pore junctions.

Analysis of Pore Connectivity for Amorphous CC1 and Amorphous CC3. Visualizations of the interconnected and the unconnected voids for the static models are shown in Figures 5

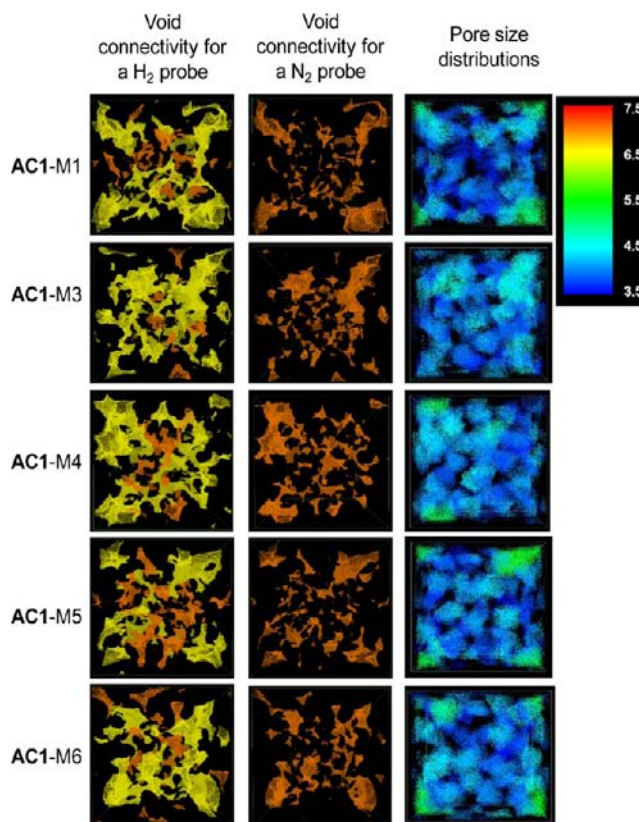


Figure 5. Visualization of void connectivity for the amorphous CC1 models. Interconnected and unconnected voids are colored yellow and orange, respectively, both for a H₂ radius of 1.42 Å (left column) and a N₂ radius of 1.82 Å (center column). The color map of pore sizes in different structural models, ranging from diameter of 3.5 to 7.5 Å is shown on the right.

and 6 (see also larger images in Figure S21 and movies S1–4 for AC1-M4 and AC3-M4, Supporting Information). The H₂ void volume is mostly connected in amorphous CC1 and fully connected in amorphous CC3, leading to extensive pore networks for both materials. This rationalizes their observed H₂ uptakes at 77 K (Figure 1c,d). In contrast, none of the voids are interconnected with respect to a N₂ probe in the denser amorphous CC1 model, as illustrated by the orange,

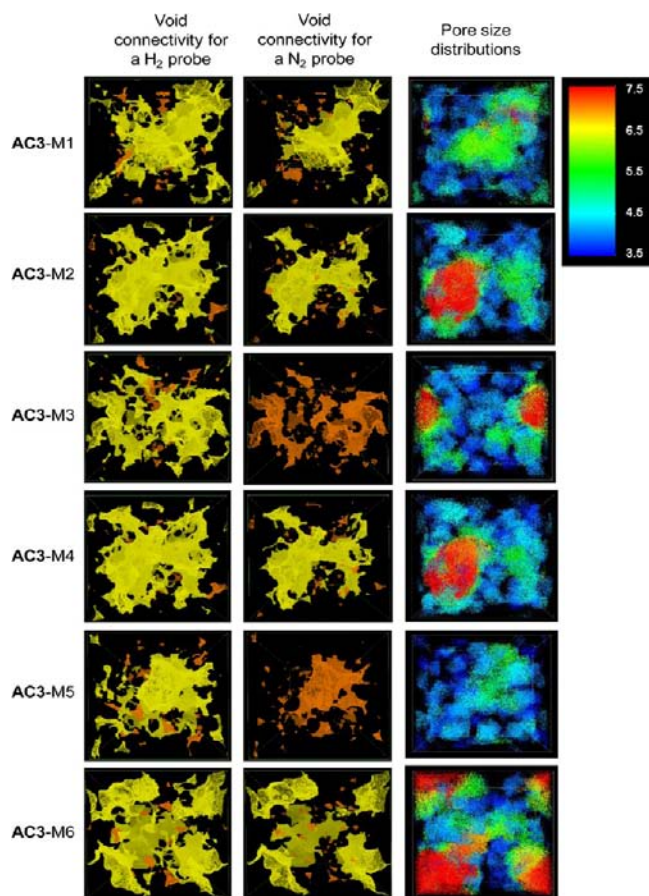


Figure 6. Visualization of void connectivity for the amorphous CC3 models. Coloring as described in Figure 5.

unconnected void space. Hence, based on this static analysis of structure, amorphous CC1 would be expected to be nonporous to N₂ and this is observed experimentally (Figure 1c). The average percentage of surface area that is ISA for the five AC1 models is 0% using a N₂ probe and 88% using a H₂ probe. The comparable values for the six AC3 models are 82% for N₂ and 97% for H₂.

Analysis of Pore Size for Amorphous CC1 and Amorphous CC3. The sizes of the voids are further mapped by color in Figures 5 and 6 (right column) using Zeo++ (see section 1.12 for details). The void sizes range from diameters of 3.5 Å (blue) to 7.5 Å (red). In the amorphous CC1 models, the majority of voids are colored blue (pore sizes ranging from 3.5–4.5 Å), which corresponds to the size of the intrinsic cage cavity (Figure S4a, Supporting Information). By contrast, voids in amorphous CC3 models are almost all colored green and red due to larger extrinsic pore cavities. This is particularly evident from a comparison across all of the models.

A purely visual analysis of void size and connectivity based on these static models suggests, correctly, that amorphous CC1 would be porous to H₂ but not to N₂, while the large number of extrinsic void cavities and predominant window-to-window packing motif in amorphous CC3 creates an interconnected pore structure that should be permeable to both H₂ and N₂, again in agreement with experiment (Figure 1). However, analysis of the ISA and void volume for static structures does not give any information on the dynamic pore connectivity, which would be expected to be important for “soft” organic

solids of this type.⁴⁵ We therefore used MD simulations to investigate gas diffusion in these amorphous solids.

3.3. Gas Diffusion Analysis. The initial configurations for gas diffusion simulations at 300 K were taken from the last configurations of the 8 ns NPT simulations at 300 K (step 3), rather than the 0 K minimized structures. Comparable structural analysis for these models was also carried out and this confirmed that there was no significant difference between structures at steps 3 and 4 (Tables S19 and S20, Supporting Information). The self-diffusion coefficients were calculated for both N₂ and H₂ in different structural models of amorphous CC1 and amorphous CC3 over a 20 ns NVT MD simulation. The cage and gas molecules were kept fully flexible. MSD plots are shown in Figure S22 and S23 (Supporting Information) and the self-diffusion coefficients in Table 4.

Table 4. Self-Diffusion Coefficients for H₂ and N₂ Diffusion in Different Amorphous CC1 and CC3 Models

self-diffusivity	D_s N ₂ (m ² s ⁻¹)	D_s H ₂ (m ² s ⁻¹)
AC1–M1	5.5×10^{-10}	1.6×10^{-8}
AC1–M3	1.4×10^{-9}	2.7×10^{-8}
AC1–M4	7.9×10^{-10}	1.7×10^{-8}
AC1–M5	5.5×10^{-10}	1.0×10^{-8}
AC1–M6	6.6×10^{-10}	1.4×10^{-8}
avg	7.9×10^{-10}	1.7×10^{-8}
AC3–M1	2.8×10^{-9}	1.1×10^{-7}
AC3–M2	6.4×10^{-9}	6.8×10^{-8}
AC3–M3	1.4×10^{-9}	6.4×10^{-8}
AC3–M4	6.7×10^{-9}	8.2×10^{-8}
AC3–M5	2.1×10^{-9}	4.3×10^{-8}
AC3–M6	2.1×10^{-9}	1.3×10^{-7}
avg	3.6×10^{-9}	8.3×10^{-8}

An average self-diffusion coefficient of 8.3×10^{-8} m² s⁻¹ over six structural models was calculated for H₂ diffusion in CC3, with a range of 4.3×10^{-8} to 1.3×10^{-7} m² s⁻¹. The different self-diffusion coefficients are due to the different topologies of the pore volume in these various models. The interconnected surfaces for the amorphous CC3 models (Figure 6), show some large cavities that allow faster H₂ diffusion and also narrower channels or small cavities that would slow down H₂ diffusion. By contrast, the H₂-accessible voids in CC1 structural models are in general smaller (Figure 5), and hence the average self-diffusion of H₂ in CC1 over the five representative structural models is about four times lower, with an average coefficient of 1.7×10^{-8} m² s⁻¹. The average self-diffusion coefficient for N₂ was calculated as 7.9×10^{-10} m² s⁻¹ for N₂ in amorphous CC1 and 3.6×10^{-9} m² s⁻¹ in amorphous CC3. The low self-diffusion coefficients for N₂ in AC3–M3 and AC3–M5 are consistent with the disconnected pore volumes observed in the static structures for these models, as shown in Figure 6. Overall, the gas diffusion in amorphous CC1 is substantially slower than in amorphous CC3.

It has been suggested that pore size and pore shape have a significant effect on the gas diffusion in porous solids.⁶⁷ Here, the range of self-diffusivity found for H₂ and N₂ in the different structural models may be due to the different topologies and connectivity of the pore volumes in the models. A comparison of self-diffusion coefficients and pore volumes is shown in Table S21 (Supporting Information) for various different types of porous materials. The H₂ and N₂ diffusion in amorphous CC1 and CC3 are slower than in MOF-5, which has a much

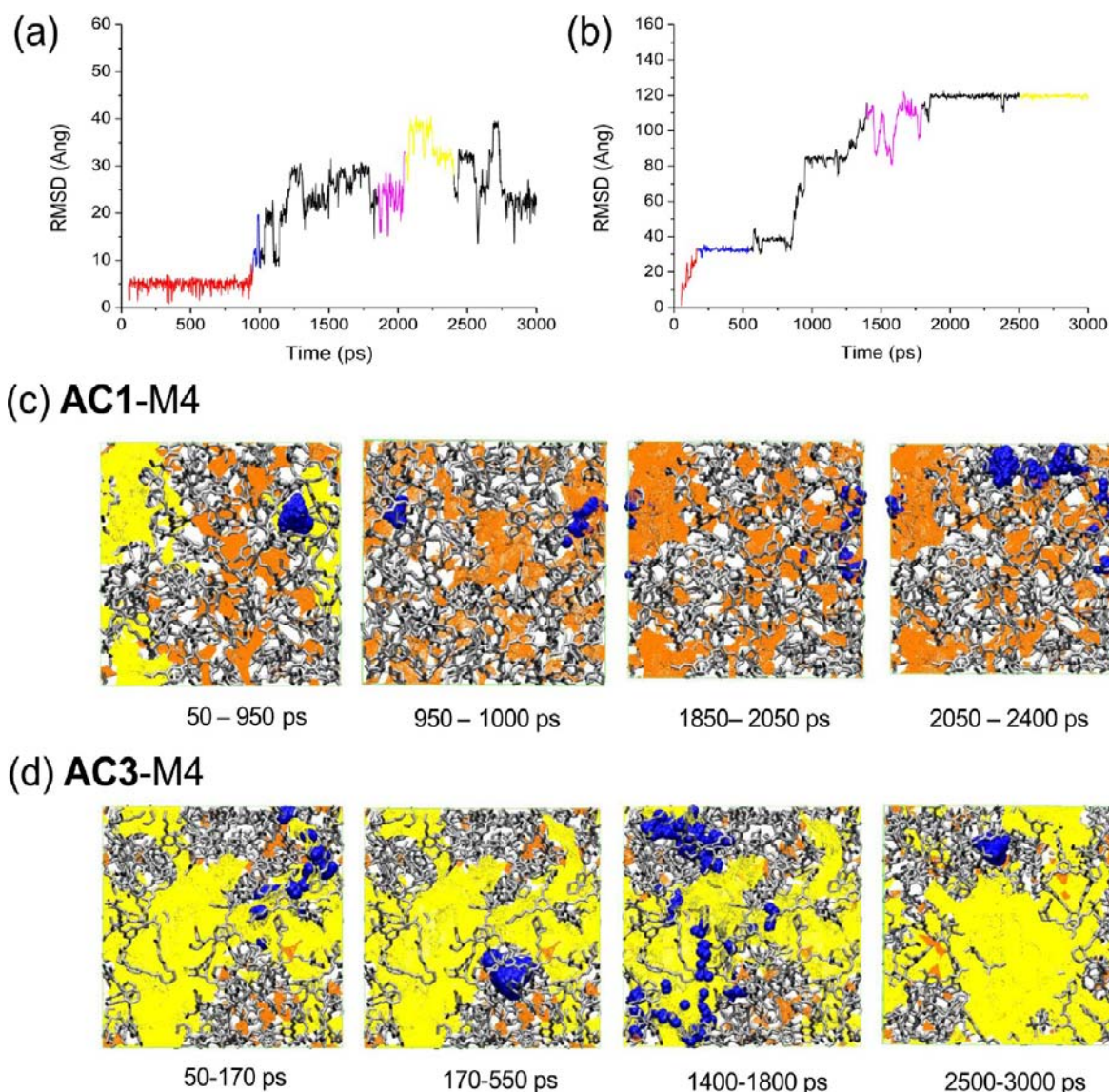


Figure 7. RMSD plots for the N₂ in (a) AC1-M4 and (b) AC3-M4. The simulation periods chosen for the analysis are highlighted in red, blue, pink, and yellow. (c) N₂ diffusion trajectories for AC1-M4 at specified time intervals. (d) N₂ diffusion trajectories for AC3-M4 at specified time intervals. The void accessibility for a 1.82 Å N₂ probe is recalculated for each of the static structures at the start of each displayed time interval. Accessible and inaccessible pore voids are colored yellow and orange, respectively.

larger pore volume, but are comparable with ZIF-68 and ZIF-70.^{68,69}

The root-mean-square displacement (RMSD) was also calculated for N₂ molecules diffusing through all of the amorphous cage models. N₂ diffusion was restricted to a limited zone for the five different amorphous CC1 models, with an average displacement distance of 20.8 Å, as shown in Figure S24 (Supporting Information). The six different amorphous CC3 models illustrate that the N₂ molecule can diffuse through a much broader zone than for CC1, with an average displacement distance of 54.3 Å in Figure S25 (for detailed analyses, see Figures S26–S29, Supporting Information).

We examine the RMSD plots in detail for two representative models, AC1-M4 and AC3-M4 in Figure 7. Rather than giving a narrative for the whole 3 ns simulation period, we highlight certain key events in the diffusion process to illustrate the diffusion mechanism. For each interval of simulation time shown in Figure 7, the pore connectivity was recalculated in Zeo++ at the beginning of that interval in order to give an

interval-relevant picture of pore connectivity at that point. Hence, for the MD snapshots in Figure 7, the pore connectivity evolves over the time scale of the gas molecule diffusing through the solid, although not necessarily in a causal sense: that is, we have no way at present of establishing whether there is a correlation between gas “hopping” and pore channels opening up in a cooperative way.

For AC1-M4, the N₂ molecule was located inside a single cage molecule in the simulation period 50–950 ps, which results in a constant average RMSD value (the initial horizontal plateau in Figure 7a, colored red). Both the interconnected and nonconnected N₂-diameter voids, as calculated at a simulation time of 50 ps, are shown in Figure 7c. Despite being located in a pocket of pore volume that is at least transiently interconnected at 50 ps (colored yellow in Figure 7c, left), the N₂ molecule does not initially leave its starting cage in the 50–950 ps period. At around 950 ps, the N₂ molecule diffuses from the intrinsic cage volume into the extrinsic void volume via one of the cage windows, which leads to an increase in the

RMSD (the first vertical step in Figure 7a, colored blue). The pore topology was recalculated for the structural snapshot at 950 ps, showing that the pore volume that was previously interconnected at 50 ps is now unconnected, as illustrated by an absence of yellow voids in the 950 ps structure in Figure 7c. Hence, while the N_2 molecule has left its starting cage, it is now located in pore volume that has become unconnected at 950 ps, and the pore volume in AC1–M4 remains largely unconnected throughout the remainder of the simulation, as indicated by subsequent snapshots in Figure 7c. The fluctuation of RMSD values between 1850 and 2050 ps (highlighted in pink in Figure 7a) is due to the N_2 molecule diffusing within extrinsic voids. These voids were technically inaccessible for a N_2 probe for the structural snapshot at 1850 ps (Figure 7c), but based on the motion of the host, these voids are accessible at points during that period. In the 2050–2400 ps simulation period (highlighted yellow), intercage “hops” occurred, whereby the N_2 molecule diffused, multiple times, between the internal void volumes of two adjacent cage molecules. At 2400 ps, the N_2 molecule diffused back into extrinsic void space. The structure at 2050 ps and the overlaid N_2 trajectory (Figure 7c, right) shows how the N_2 molecule has been able to traverse void space that is formally disconnected in the 2050 ps snapshot by virtue of the dynamic nature of the pore structure, where unconnected pore space can become transiently connected. This would not be expected from analysis of static snapshots alone. To our knowledge, this is the first example of the molecular simulation of dynamic guest diffusion in an amorphous, porous molecular solid. The amount of diffusion for N_2 is, however, still very limited and is insufficient to be manifested as N_2 adsorption in real bulk samples of amorphous CC1 at 77 K (Figure 1c).

A quite different picture emerges for amorphous CC3. In AC3–M4, the N_2 molecule diffuses from inside its starting cage to the large, interconnected extrinsic pore cavities in the 50–170 ps simulation time interval, resulting in an increased RMSD value (period highlighted in red in Figure 7b). At 50 ps, and throughout the rest of the simulation, the AC3–M4 pore structure is largely interconnected, as evidenced by the predominance of yellow-colored channels in all interval snapshots in Figure 7d. The N_2 diffusion trajectory then follows these interconnected void volumes. The two plateaus highlighted in blue and yellow in Figure 7b correspond to periods when the N_2 molecule is located in a single cage cavity. As for CC1, the simulations show how the N_2 trajectory can traverse formally unconnected pore volumes via dynamic mechanisms and transient pore connectivity, although the CC3 structure has much less of this unconnected pore volume to traverse.

3.4. Cage Occupancy and Gas Hopping Analysis for Amorphous CC1 and CC3. To further rationalize the gas selectivity and to understand gas diffusion in these amorphous systems, we analyzed the cumulative number of cages occupied and the number of “hopping” events where a gas moved between a pair of cages. Again, one representative model for both amorphous CC1 (AC1–M4) and CC3 (AC3–M4) was chosen for this analysis, picked on the basis of being the closest to the average self-diffusion coefficient for the combined models. Figure 8 demonstrates the cumulative number of cages occupied in the simulations. Two different starting positions for N_2 and H_2 were chosen in each of the models for the cumulative cage occupancy analysis in order to evaluate the sensitivity of the resulting trajectory to the starting position.

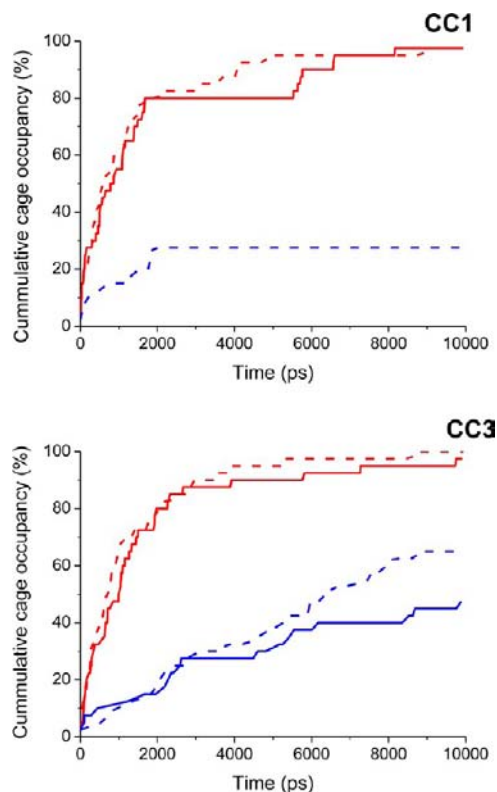


Figure 8. Cumulative cage occupancy during the 10 ns MD simulation for H_2 (red) and N_2 (blue) in (top) amorphous CC1 and (bottom) CC3. For N_2 , the molecule was placed at accessible voids (solid lines) and at inaccessible voids (dashed lines).

The starting positions corresponded to voids that were interconnected and unconnected at the start of the simulation, although only unconnected voids were observed for N_2 in amorphous CC1, and hence only one starting position was evaluated for this combination.

As Figure 8 shows, essentially all cages in the simulation cell are visited by H_2 in the 10 ns simulation for both amorphous CC1 and CC3. The time taken for the H_2 molecule to visit 39 cages was approximately 8–9 ns in amorphous CC1, dependent upon the starting position. For amorphous CC3, the equivalent time was either 5 or 10 ns depending on the starting position. This indicates, qualitatively, that the rate of H_2 diffusion is fairly comparable in both amorphous CC1 and amorphous CC3. The situation with N_2 is quite different: it did not occupy all 40 cages during a 10 ns simulation, either for amorphous CC1 or CC3. Occupancy of 27.5% of the available cages was observed for N_2 diffusion in amorphous CC1 in this period, as compared to either 45% or 65% for amorphous CC3, dependent upon the starting position. These results, which probe the percentage of the available cage volumes traversed in a 10 ns period, are consistent with the faster self-diffusion coefficients calculated for N_2 in CC3.

Trajectories of a H_2 and N_2 molecule for the entire 10 ns simulation are shown in Figure 9 for the two models, showing that H_2 has a broader range of diffusion in the solid compared with N_2 , especially for amorphous CC1 where N_2 diffusion occurs only in a limited region (Figure 9b). The hopping analysis, as detailed in sections 12 and Table S18 of the Supporting Information, shows that the average interval between cage hops for H_2 is similar in both of the amorphous cage systems (11–20 ps). By contrast, the average interval

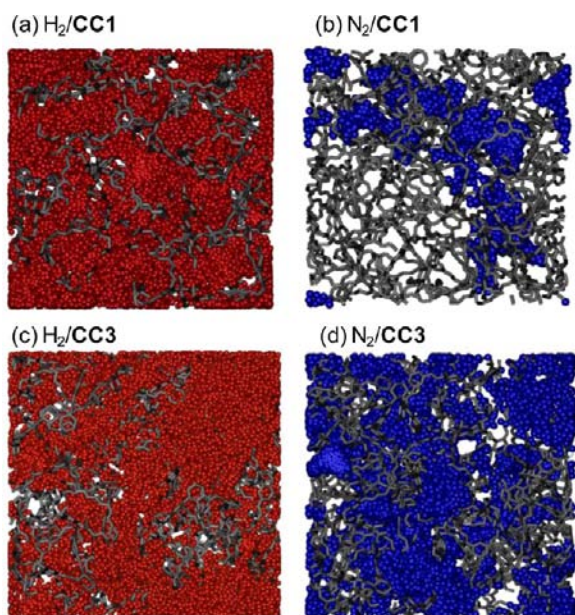


Figure 9. Complete diffusion trajectories of the gas molecules over the 10 ns simulations, sampled every 0.5 ps: (a) H_2 and (b) N_2 in amorphous CC1; (c) H_2 and (d) N_2 in amorphous CC3.

between cage hops for N_2 in amorphous CC1 is 250 ps, about 2–3 times longer than for amorphous CC3 (80–100 ps). This molecular insight into diffusion mechanisms can, at present, only be obtained by simulations, and these studies give an unprecedented level of detail regarding diffusion phenomena in solids of this type. We refer the reader to a more in depth analysis of gas hopping events in sections 11 and 12.

4. CONCLUSIONS

We describe a novel computational methodology for modeling the solid-state, amorphous packing of porous organic cages, for analyzing void connectivity, and for simulating the diffusion of gases within the pore structure of these materials. Good agreement between data derived from models and from experiment, including gas sorption and X-ray analysis, suggests that the structural models are representative and can give us a molecular level understanding that is unobtainable by any means other than simulations. Our future goal is to use these *in silico* methods to design new porous molecules with properties in the amorphous state that are tailored for applications.

Several conclusions can be drawn from these simulations. First, the high level of porosity in amorphous CC3 with respect to its crystalline form, as observed experimentally,¹⁴ is explained by the molecular packing in these amorphous CC3 models. These models also provide insights into the underlying reasons for this, over and above the somewhat obvious inference that less efficient packing in the amorphous solid leads to lower density and additional extrinsic pore volume that is not present in the denser, crystalline state. In particular, the interconnectivity of this additional extrinsic pore volume is enhanced by a statistical preference for CC3 to form intermolecular window-to-window interactions in the amorphous state (Figure 4), which in turn stems from the cyclohexane vertices on CC3 and its aspherical shape. The windows therefore have a tendency to “interlock”, as also observed in the crystalline form of CC3.⁴⁷ This window-to-window preference does not occur in the more spherical cage analogue CC1 and this, coupled with the higher

density that arises in the simulations, leads to a much less interconnected pore structure and significantly smaller pore sizes in the amorphous state. The combination of these factors explains the observed selectivity for H_2 over N_2 for amorphous CC1 (Figure 1), even if one only considers static pictures of the available pore volume (Figure 5–6).

Our simulations also show, however, that the diffusion of gases within these amorphous solids cannot be wholly understood from static pictures and that the interconnectivity of the pore channels fluctuates over time for these amorphous solids. For the first time, we have used the Zeo++ tool⁶² to begin to analyze the dynamic evolution of pore connectivity, as shown in Figure 7. These MD simulations reveal that N_2 diffusion is indeed restricted in amorphous CC1 but that H_2 can diffuse through the whole simulation cell, which rationalized the experimentally observed ideal H_2/N_2 gas selectivity of 19 (Figure 1). Furthermore, N_2 and H_2 diffusion in amorphous CC3 is calculated to be faster than in amorphous CC1 due to its larger interconnected pore volume and larger pore sizes. However, even in amorphous CC1, these MD simulations show that pore volume that is unconnected with respect to a particular guest can become transiently interconnected as a result of molecular motion in the solid (e.g., Figure 7c, left). This offers a direct explanation for the phenomenon of “porosity without pores”⁷⁰ that is also known for crystalline molecular solids, such as calixarenes.⁷¹ We believe that this is the first example of a molecular dynamics simulation for an amorphous molecular solid that provides a direct visualization of this dynamic gating process in action.

In principle, amorphous molecular solids might lead to new functional materials that cannot easily be obtained with insoluble extended frameworks. For example, we envisage functional porous amorphous coatings, created via codeposition from solution of porous cage molecules along with a molecular catalyst to create a “ship-in-a-bottle” catalyst system. Likewise, soluble cages might be deposited as amorphous coatings to form gas separation barrier layers. Molecular simulations of properties such as pore size, pore connectivity, and diffusion selectivity could allow us in the future to carry out *de novo*, *in silico* design of amorphous porous solids, paralleling recent developments in the computational prediction of structure,^{72,73} thermodynamic stability, and physical properties for crystalline porous solids.^{48,74}

■ ASSOCIATED CONTENT

📄 Supporting Information

Computational details, experimental PXRD, BET, total scattering data, simulated pair disruption functions and total scattering data, pore size distributions, calculated pore sizes, surface areas, pore volumes, gas diffusion, and gas diffusion behavior analysis, X-ray data (CIF). Four video files showing void connectivity in the representative models AC1–M4 and AC3–M4. This information is available free of charge via the Internet at <http://pubs.acs.org/>.

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Notes

The authors declare no competing financial interest.

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