

Supplementary Information

1 Detailed Methods

The all-silica zeolites frameworks were selected from the 110 orthogonal structures in the International Zeolite Association (IZA) database¹ and the orthogonal structures in the predicted crystallography open database (PCOD)². The PCOD database² was reduced to a set of 121,966 structures by removing those with a largest free-sphere diameter below 2.25 Å as these are less accessible to carbon dioxide (CO₂) (Figs. S13–S14); of these, a randomly chosen subset of 81,526 structures was screened. Note that some of the results for the IZA zeolites appear outside the range of the hypothetical zeolite results, which may be due to the energy minimization step used in the construction of the hypothetical zeolite database.² The geometric structure descriptors of largest included sphere diameter, largest free sphere diameter, and accessible surface area was obtained using the open-source software Zeo++,³ using its high accuracy setting,⁴ a methane (CH₄) probe of 1.625 Å radius, a silicon atom radius of 2.10 Å, and an oxygen atom radius of 1.52 Å.

Mixture adsorption isotherms were produced both directly by grand-canonical Monte Carlo (GCMC) simulation⁵ and from the simulated pure-component isotherms using Ideal Adsorbed Solution Theory (IAST).⁶ Except for where the two results were compared, the directly-generated mixture isotherms were used throughout this paper. All isotherms were generated with a GPU GCMC code that has been described elsewhere,^{7,8} which uses a parallel flood fill algorithm to find blocked pockets inaccessible from the gas phase⁷ and uses density-biased sampling to accelerate convergence.⁸ The force field developed by García-Pérez *et al.*⁹ was used, which consists of Lennard-Jones and Coulombic terms for guest-guest and guest-host interactions. The host framework atoms were assumed to be rigid, and the number of simulated unit cells was chosen such that the simulation box extended at least twice the cutoff radius of 12 Å. Our work exclusively used units of absolute loading.

To directly generate mixture isotherms, the GPU GCMC code was adapted to allow for multiple adsorbate species, and blocked pockets were separately found and applied to CH₄ and CO₂. For all zeolites, the number of equilibration and production steps, respectively, were set to 10,000,000 and 1,000,000 for the associated petroleum gas (APG) and non-associated gas (NAG) processes for which isotherms were simulated up to 100 bar, and 5,000,000 and 1,000,000 for the landfill gas (LFG) process for which isotherms were simulated up to 5 bar. These numbers of steps were validated as being sufficient for isotherm convergence⁸ at the pressures involved for screening the three processes studied in this paper by comparison to isotherms created using a conventional CPU-based code (Figs. S24–S29).

To generate mixture isotherms by IAST, pure-component isotherms for the IZA zeolites were generated using 5,000,000 equilibration and 3,000,000 production steps, pure-component isotherms for the hypothetical zeolites with a largest free-sphere diameter less than 3.75 Å were generated using 1,250,000 equilibration and 500,000 production steps, and pure-component isotherms for the hypothetical zeolites with a largest free-sphere diameter greater than 3.75 Å were taken from previous work.^{10,11} The number of steps used to generate pure-component isotherms for the hypothetical zeolites with a largest free-sphere diameter less than 3.75 Å were chosen to achieve a similar degree of convergence as the pure-component isotherms taken from previous work. The pure-component isotherms were then fit via the method of least squares to single- or dual-site Langmuir isotherms for each adsorbate based on which fit gave a larger adjusted R² value for that adsorbate. The spreading pressure was then calculated via analytical integration of equation 19 of Myers and Prausnitz⁶. As this procedure sometimes requires the fitted isotherms to be evaluated at pressures above the highest pressure point of the simulated isotherms, we only allowed such extrapolation for structures which had fitted saturation loading capacities less than ten times the amount of the loading at the highest simulated pressure

point; the remainder were not used in comparing IAST results to directly-generated mixture isotherm results. The fitted saturation loading capacities of these pure-component adsorption isotherms were used in plots in this paper as stated (fitted Henry coefficients were not used in plots as we instead used values obtained from Widom insertions; see below). When a dual-site Langmuir isotherm was used, the sum of the two saturation loading capacities was used.

The Peng-Robinson equation of state¹² was used to convert between pressure and fugacity, with the critical temperatures, critical pressures, and acentric factors of CH₄ and CO₂ being 190.6 K, 46.0 bar, 0.008, 304.2 K, 73.76 bar, and 0.225, respectively. For mixture isotherms generated directly by GCMC, the known total pressure and mole fractions were converted to component fugacities using the van der Waals mixing rules with a binary interaction parameter of 0.0919, which were then used as inputs to the simulations. For pure-component isotherms used for IAST, the component’s fugacity was input directly to the simulation and then converted to pressure prior to IAST calculations; however, in Fig. 8 fugacity was not converted to pressure prior to the IAST calculation to ensure consistency of units on the x-axis.

Widom insertions were performed by the GPU code^{7,8} to calculate isosteric heats of adsorption, Henry coefficients, and helium void fractions, all at 300 K. Some zeolites have no enthalpically favorable adsorption sites for CH₄, either intrinsically or due to blocking; Q_{st,CH_4} for these materials was set to 0 kJ mol⁻¹ for the purposes of plotting and for calculating the objective function. Isosteric heats of adsorption and Henry coefficients for all materials were calculated using 200,000 Widom insertions. For all materials, helium void fractions were calculated using 100,000 Widom insertions and the force field of Talu and Myers¹³.

As different materials perform optimally at different conditions, we evaluated each material at its own optimal desorption conditions to allow for fair comparison. For each zeolite, the optimal desorption pressure (PSA), temperature (TSA), or both (PTSA) was found by minimization of the objective function value (OFV). We note that future studies may need not strictly find each adsorbent’s optimal desorption conditions if desorption conditions can be well-approximated prior to the study’s commencement, as we found that the Separation Performance Parameters (SPP values) for the adsorbents undergoing the three PSA processes at set desorption pressures are well-correlated with the SPP values for the adsorbents undergoing the three PSA processes at each adsorbent’s optimal desorption pressure (Fig. S30). For our study, we used the optimal desorption conditions for all analyses except for the data presented in Fig. S30 and for the comparison of the various metrics presented in Figs. 1 and S1, since we found that the Sorbent Selection Parameter and Adsorbent Performance Indicator could not be used to find the optimal desorption conditions. SciPy version 0.15.1 was used to perform the minimization.¹⁴ The adsorption pressures were set by the process, and the adsorption temperature was always 300 K. The desorption temperature was constrained to be greater than or equal to 300 K for TSA and PTSA and set to 300 K for PSA, while the desorption pressure was constrained to be between 0.1 bar and 1 bar for PSA and PTSA and set to 1 bar for TSA. The mole fractions were set by the process, and the same mole fraction was used for both adsorption and desorption. For all materials, mixture isotherms generated at 300 K were used, and for the IZA zeolites, upon which TSA and PTSA were also performed, mixture isotherms were also generated at higher temperatures going up in 5 K increments. Continuous optimization was applied to find the optimal desorption pressure, applying Akima cubic spline interpolation¹⁵ as implemented by SciPy¹⁴ to calculate loadings between the pressures that were simulated. Discrete optimization was applied to find the optimal desorption temperature in 5 K increments. For PTSA, pressure optimization was performed for all desorption temperatures, and the temperature that gave the lowest OFV was then chosen. When mixture isotherms gave a negative working capacity for either adsorbate (possible since mixture isotherms are not necessarily monotonically increasing functions), the working capacity for that adsorbate was set to 0 mol kg⁻¹ prior to calculation of the metrics that go into the objective function.

To implement the random forest of decision trees regression algorithm and calculate the importances of the individual descriptors, we used open-source scikit-learn version 0.14.1-2.¹⁶ Our forest included 1,000 trees, which has been shown to be an adequate number for random forest accuracy.¹⁷ Nodes were expanded until all leaves were pure.

We have placed the computer code we developed for the screening online with the Open Science Framework, along with the pure and mixture isotherms and the screening results.¹⁸ The code is capable of computing the SPP, Sorbent Selection Parameter, or Adsorbent Performance Indicator at optimized or set

desorption conditions. Mixture isotherms can be input directly as loadings at multiple pressures, in which case the code will perform interpolation to calculate loadings, or mixture isotherms will be calculated by the code using IAST with pure-component isotherms input either as loadings at multiple pressures or as already-fitted dual-site Langmuir isotherms.

2 Derivation of Metrics

The derivation of the terms in the SPP begins by calculating the mole fraction of CH₄ in the raffinate stream:

$$\begin{aligned}
 y_{\text{CH}_4,\text{raff}} &= \frac{M_{\text{CH}_4,\text{raff}}}{M_{\text{CH}_4,\text{raff}} + M_{\text{CO}_2,\text{raff}}} \\
 &= \frac{M_{\text{CH}_4,\text{feed}} - \Delta q_{\text{CH}_4} M_{\text{ads}}}{(M_{\text{CH}_4,\text{feed}} - \Delta q_{\text{CH}_4} M_{\text{ads}}) + (M_{\text{CO}_2,\text{feed}} - \Delta q_{\text{CO}_2} M_{\text{ads}})} \\
 &= \frac{1 - \Delta q_{\text{CH}_4} \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}}}{(1 - \Delta q_{\text{CH}_4} \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}}) + (\frac{y_{\text{CO}_2,\text{feed}}}{y_{\text{CH}_4,\text{feed}}} - \Delta q_{\text{CO}_2} \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}})} \\
 &= \frac{1 - \Delta q_{\text{CH}_4} \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}}}{\frac{1}{y_{\text{CH}_4,\text{feed}}} - \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}} (\Delta q_{\text{CH}_4} + \Delta q_{\text{CO}_2})}
 \end{aligned}$$

Rearranging, we obtain:

$$\frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}} = \frac{y_{\text{CH}_4,\text{raff}} - y_{\text{CH}_4,\text{feed}}}{y_{\text{CH}_4,\text{feed}} [y_{\text{CH}_4,\text{raff}} (\Delta q_{\text{CH}_4} + \Delta q_{\text{CO}_2}) - \Delta q_{\text{CH}_4}]} \quad (1)$$

We then calculate the moles of recovered CH₄ per moles of feed CH₄:

$$\begin{aligned}
 \frac{M_{\text{CH}_4,\text{raff}}}{M_{\text{CH}_4,\text{feed}}} &= \frac{M_{\text{CH}_4,\text{feed}} - \Delta q_{\text{CH}_4} M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}} \\
 &= 1 - \Delta q_{\text{CH}_4} \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}} \\
 &= 1 - \Delta q_{\text{CH}_4} \left(\frac{y_{\text{CH}_4,\text{raff}} - y_{\text{CH}_4,\text{feed}}}{y_{\text{CH}_4,\text{feed}} [y_{\text{CH}_4,\text{raff}} (\Delta q_{\text{CH}_4} + \Delta q_{\text{CO}_2}) - \Delta q_{\text{CH}_4}]} \right) \\
 &= \frac{y_{\text{CH}_4,\text{raff}} [y_{\text{CH}_4,\text{feed}} (\Delta q_{\text{CH}_4} + \Delta q_{\text{CO}_2}) - \Delta q_{\text{CH}_4}]}{y_{\text{CH}_4,\text{feed}} [y_{\text{CH}_4,\text{raff}} (\Delta q_{\text{CH}_4} + \Delta q_{\text{CO}_2}) - \Delta q_{\text{CH}_4}]} \quad (2)
 \end{aligned}$$

Finally:

$$\begin{aligned}
 \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{raff}}} &= \frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{feed}}} \div \frac{M_{\text{CH}_4,\text{raff}}}{M_{\text{CH}_4,\text{feed}}} \\
 &= \frac{y_{\text{CH}_4,\text{raff}} - y_{\text{CH}_4,\text{feed}}}{y_{\text{CH}_4,\text{raff}} [y_{\text{CH}_4,\text{feed}} (\Delta q_{\text{CH}_4} + \Delta q_{\text{CO}_2}) - \Delta q_{\text{CH}_4}]} \quad (3)
 \end{aligned}$$

Since $\frac{M_{\text{CH}_4,\text{raff}}}{M_{\text{CH}_4,\text{feed}}}$ is included as part of $\frac{M_{\text{ads}}}{M_{\text{CH}_4,\text{raff}}}$, which in turn is included in the equation for SPP, one could reasonably exclude the fractional CH₄ recovery term from appearing directly in the equation for SPP. However, a high fractional CH₄ recovery is desirable for more than direct economic costs, as any CH₄ not captured will likely be emitted and can be considered to be an environmental cost. Thus, we choose to directly include the fractional CH₄ recovery term in our equation for SPP.

3 Supplementary Figures

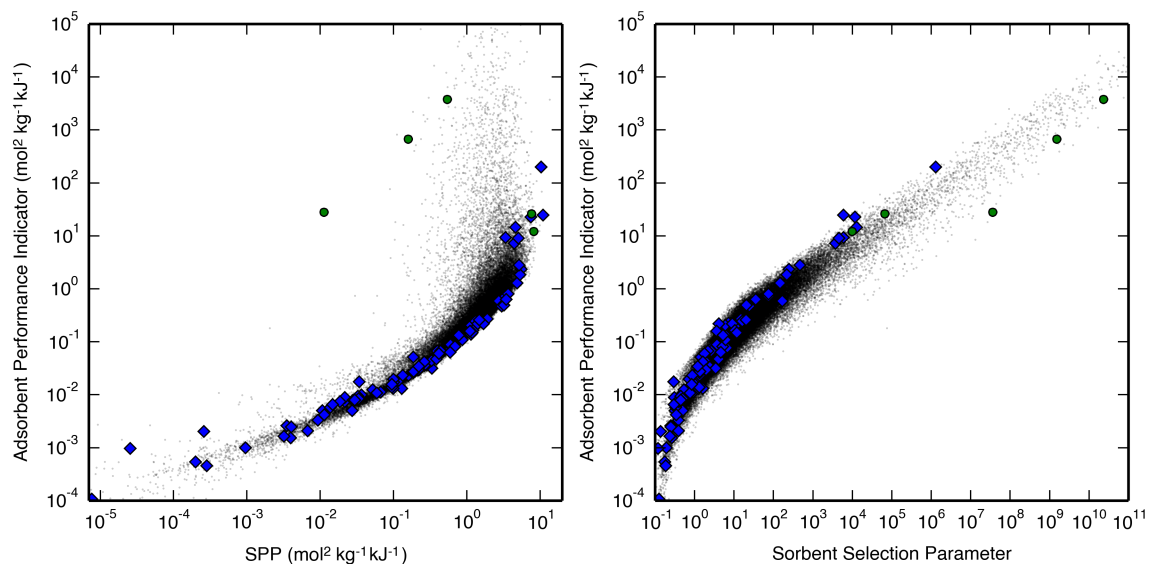


Fig. S1: (left) A correlation between the SPP and the Adsorbent Performance Indicator and (right) between the Sorbent Selection Parameter and the Adsorbent Performance Indicator, both for the APG process carried out with PSA. Hypothetical zeolites are shown as black dots, IZA zeolites are shown as blue diamonds, and the hypothetical zeolites shown in Table 1 are shown as green circles.

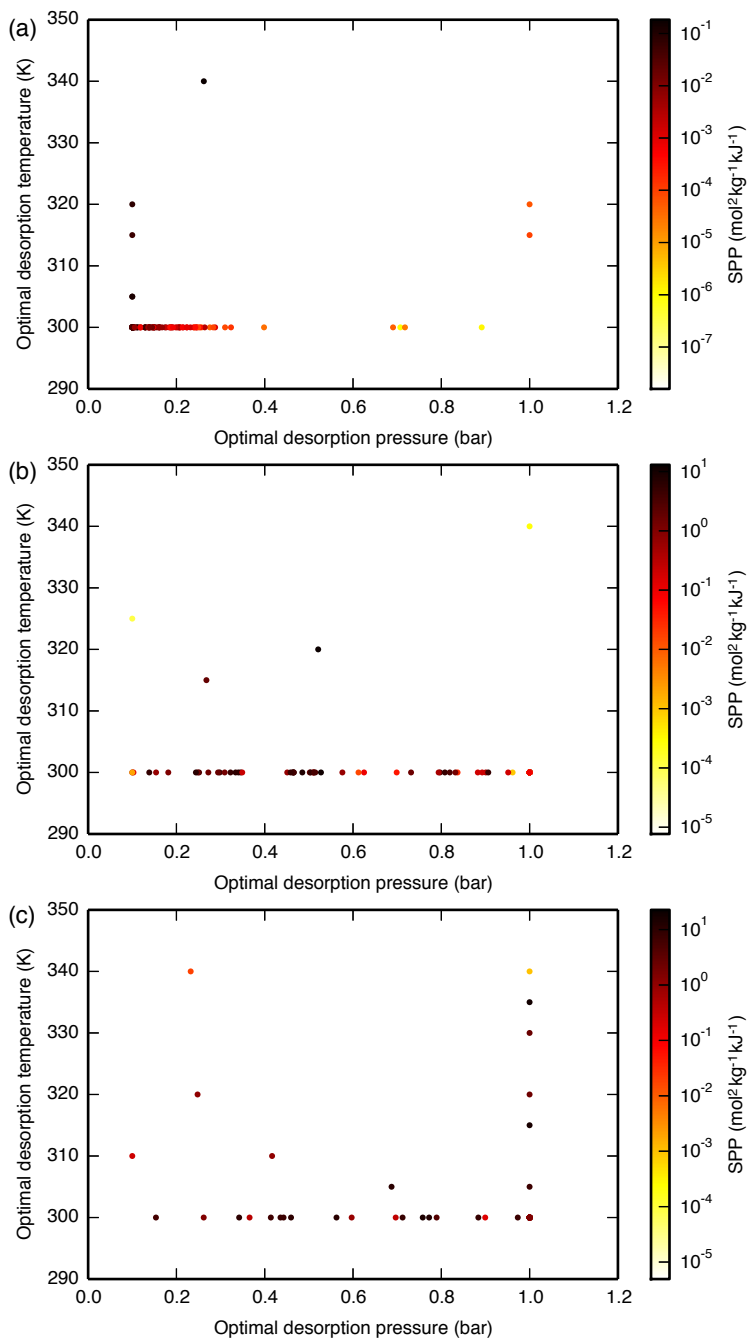


Fig. S2: SPP of the IZA zeolites for the (a) LFG, (b) APG, and (c) NAG processes carried out with PTSA as a function of the optimal desorption pressure and temperature. Note that the APG and NAG processes have several data points overlapping at 1 bar and 300 K. For the LFG process, a clear correlation exists between the desorption conditions and the SPP: zeolites with lower optimized desorption pressures are also better-performing. By comparing the SPP values of these materials at their optimal desorption conditions with the SPP values of these materials at set desorption conditions of 0.1 bar and 300 K (Fig. S30) it becomes clear that it was not the lowered desorption pressure that caused some materials to perform better than others; rather, top-performing materials have more to gain by pulling additional vacuum than do the poor-performing materials.

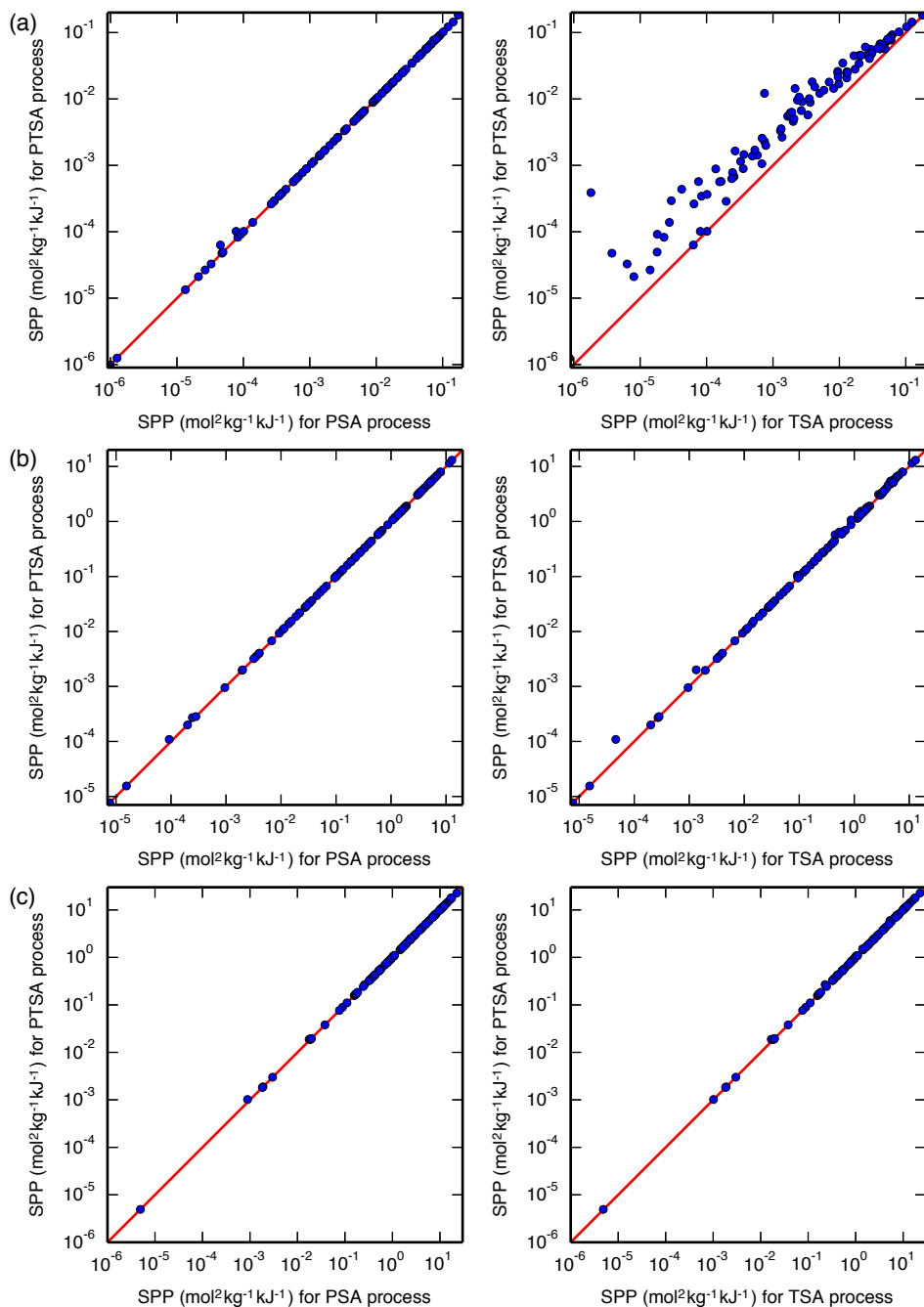


Fig. S3: SPP of the IZA zeolites for the (a) LFG, (b) APG, and (c) NAG processes carried out with PTSA at optimal desorption temperatures and pressures is plotted (left) against the SPP of the same process carried out with PSA at a set desorption temperature of 300 K and an optimal desorption pressure or (right) against the SPP of the same process carried out with TSA at an optimal desorption temperature and a set desorption pressure of 1 bar. Of the TSA processes, only the LFG process appears to have benefited from pulling vacuum, while the two higher-pressure processes received a “free” pressure-swing down to 1 bar, and thus do not benefit as much from the additional vacuum. A line is drawn at $y = x$ for reference.

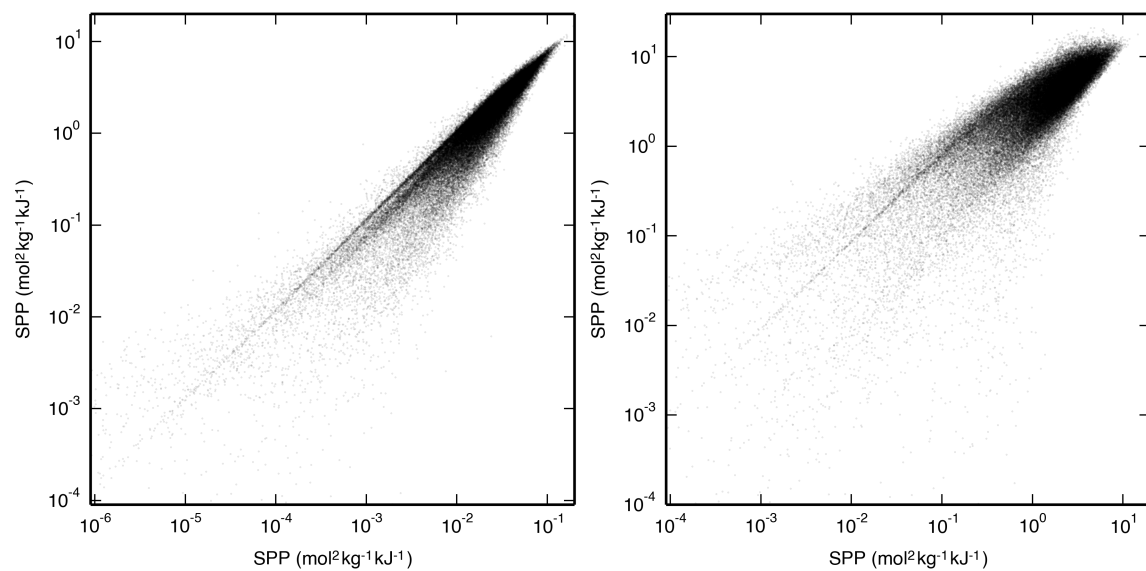


Fig. S4: SPP of the hypothetical zeolites for the (a) LFG and APG processes and for the (b) APG and NAG processes, all carried out with PSA.

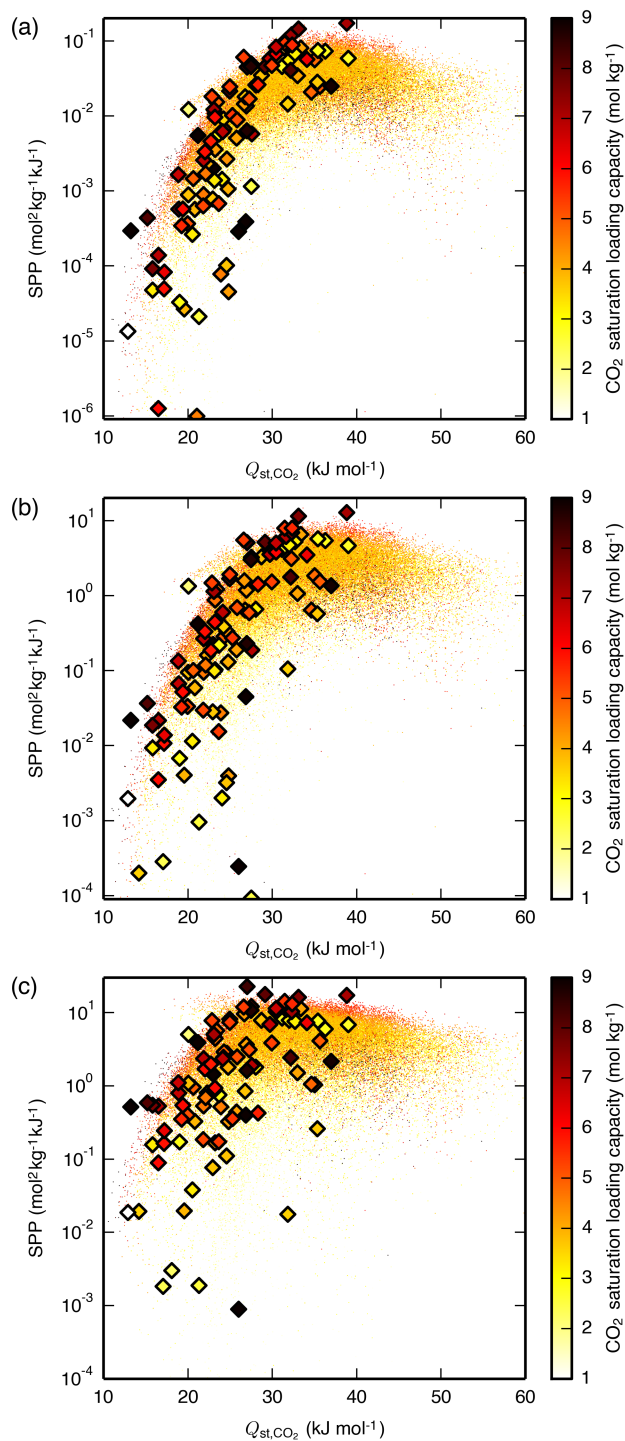


Fig. S5: SPP as a function of Q_{st,CO_2} and the CO_2 saturation loading capacity, for the (a) LFG, (b) APG, and (c) NAG processes, all carried out with PSA. The materials are plotted in random order such that the data shown are representative of the materials hidden due to having similar Q_{st,CO_2} and SPP. IZA zeolites are shown as diamonds. This plot is the same as in Fig. 2, but with the IZA zeolites overlaid.

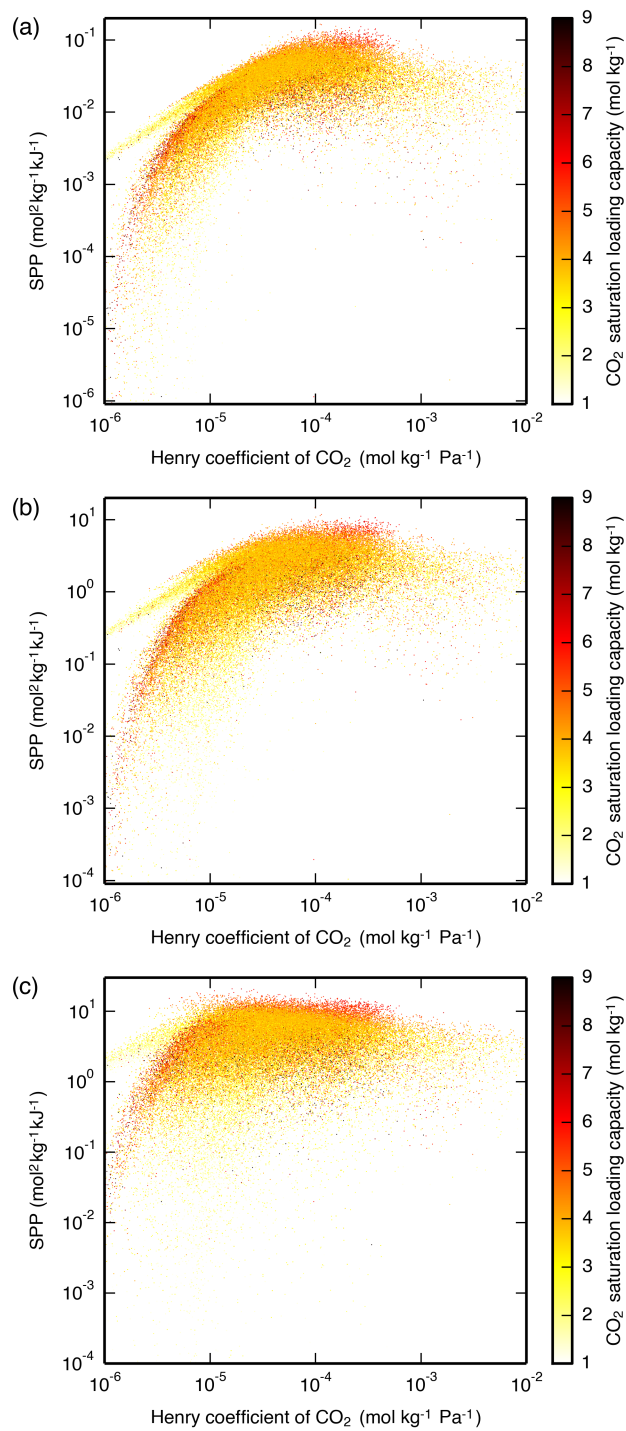


Fig. S6: SPP of the hypothetical zeolites as a function of the Henry coefficient and CO₂ saturation loading capacity, for the (a) LFG, (b) APG, and (c) NAG processes, all carried out with PSA. The materials are plotted in random order such that the data shown are representative of the materials hidden due to having similar Q_{st,CO_2} and SPP.

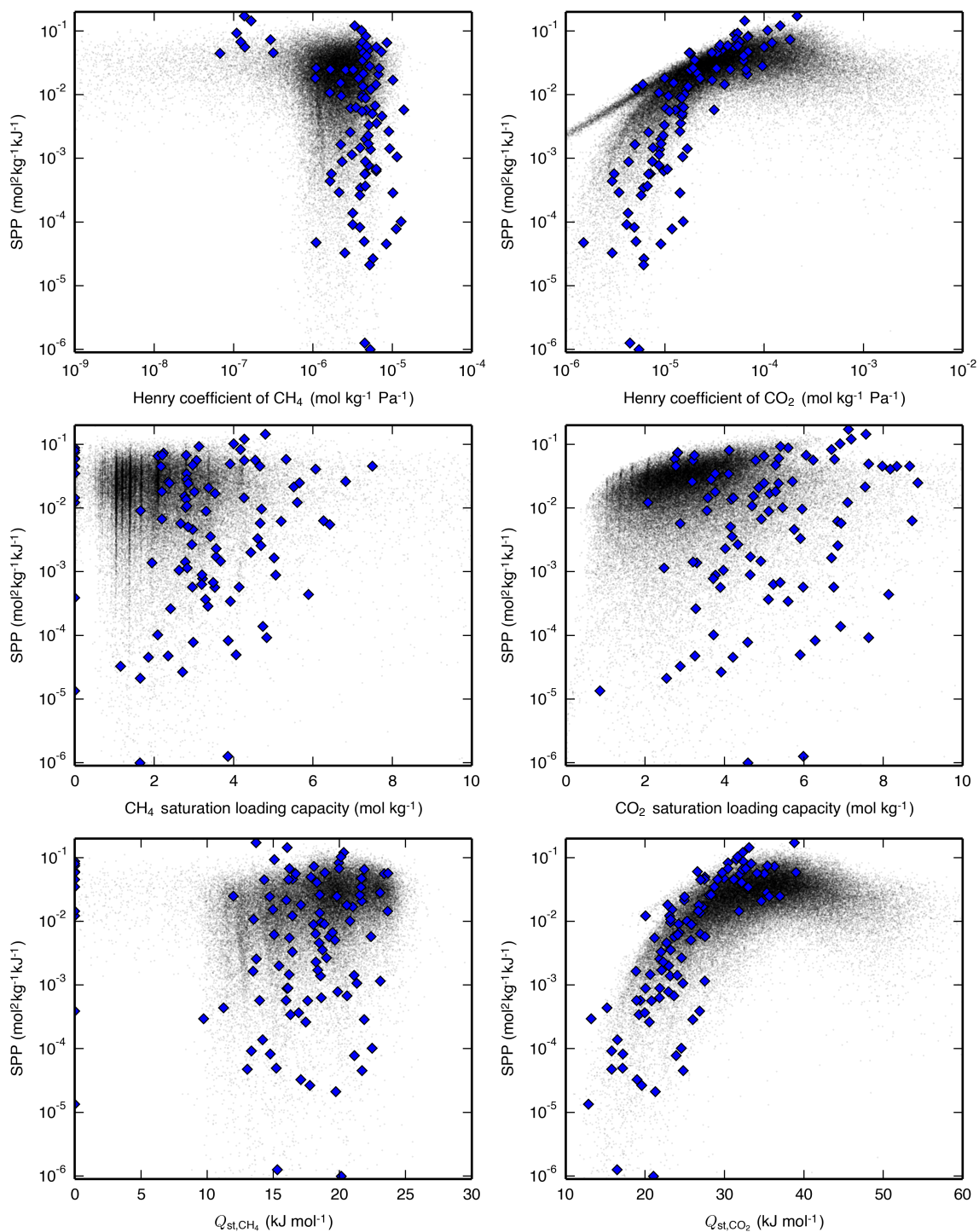


Fig. S7: SPP as a function of Henry coefficients, saturation loadings, and Q_{st} for the LFG process carried out with PSA. Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds. The stripes seen in plots containing CH₄ saturation loading capacity are due to that variable being more likely to be integer values of CH₄ molecules per unit cell (Fig. S10).

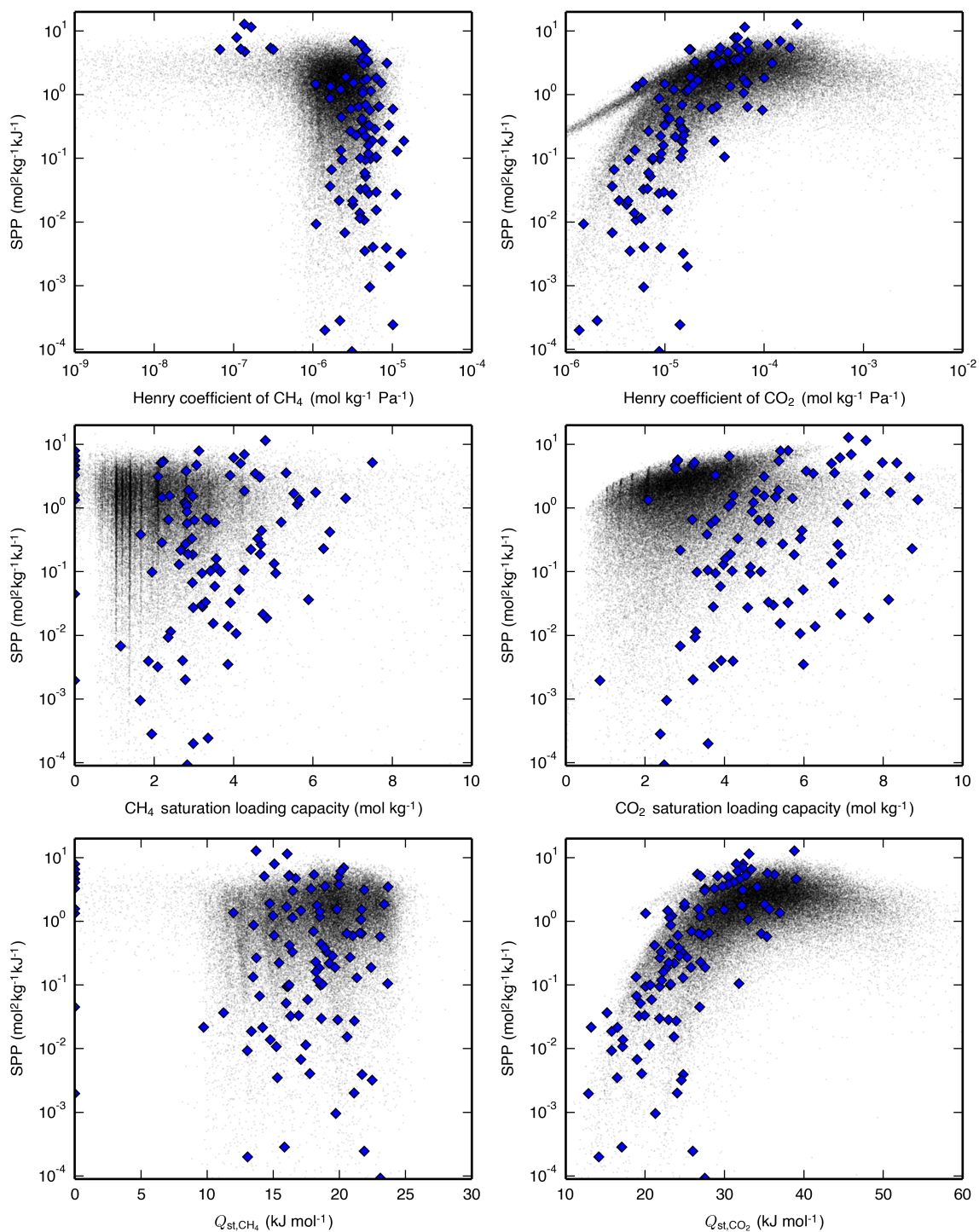


Fig. S8: SPP as a function of Henry coefficients, saturation loadings, and Q_{st} for the APG process carried out with PSA. Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds. The stripes seen in plots containing CH_4 saturation loading capacity are due to that variable being more likely to be integer values of CH_4 molecules per unit cell (Fig. S10).

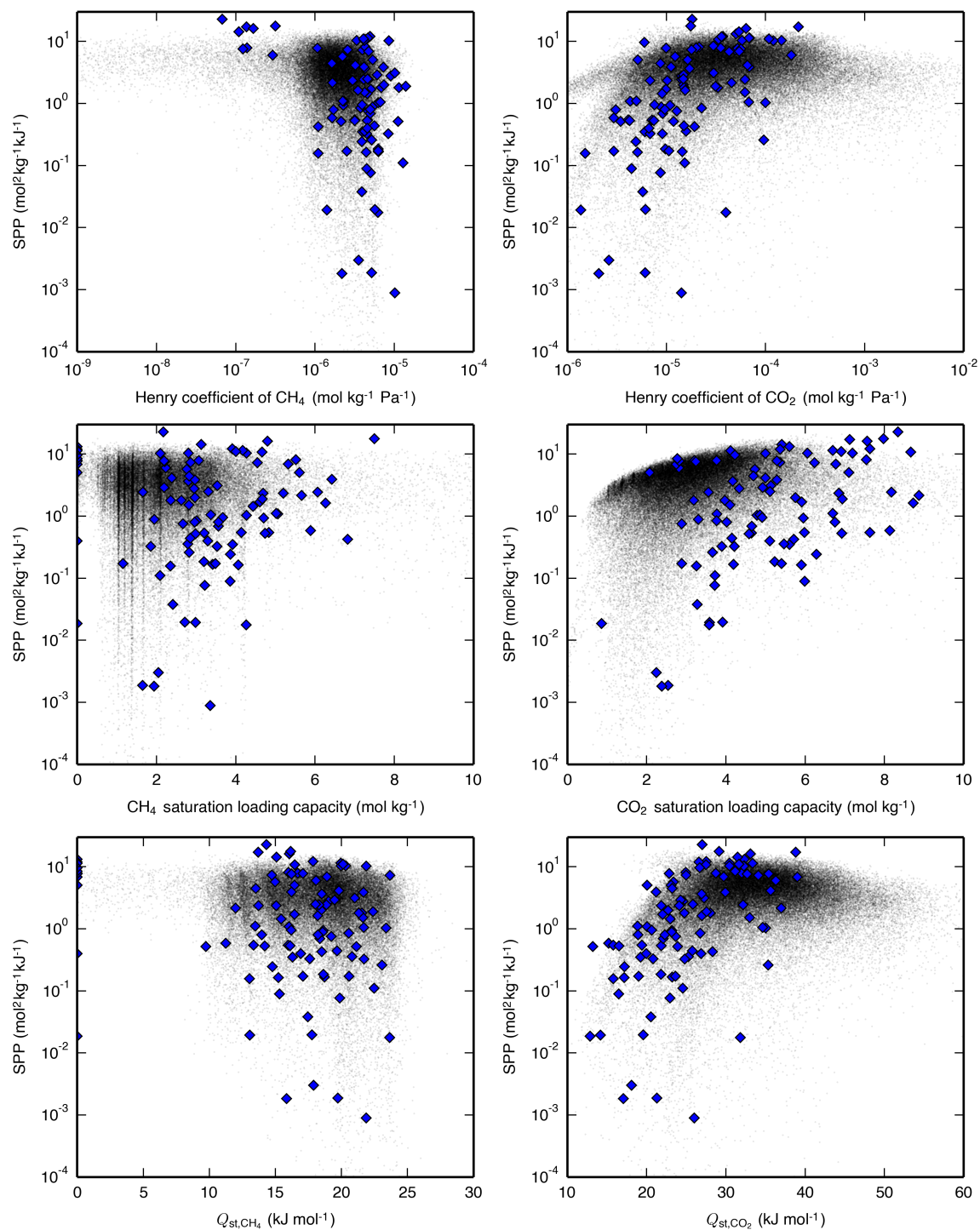


Fig. S9: SPP as a function of Henry coefficients, saturation loadings, and Q_{st} for the NAG process carried out with PSA. Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds. The stripes seen in plots containing CH_4 saturation loading capacity are due to that variable being more likely to be integer values of CH_4 molecules per unit cell (Fig. S10).

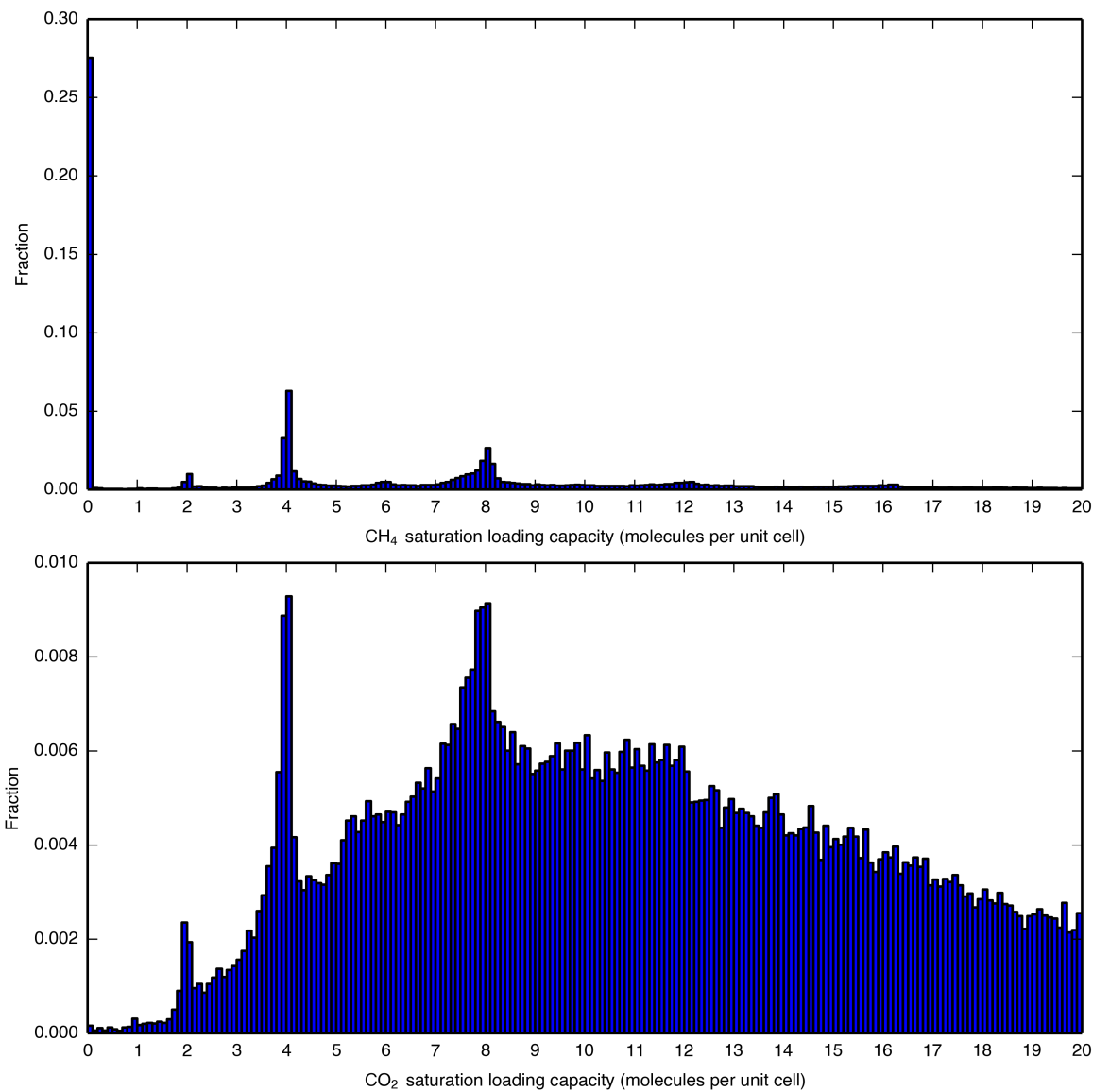


Fig. S10: The distribution of (top) CH₄ saturation loading capacities for the hypothetical zeolites shows that even integer values in units of molecules per unit cell are most common, a trend which is less noticeable for (bottom) CO₂ saturation loading capacities. This explains the stripes seen in Figs. 5, S7–S9, and S13 in the plots containing CH₄ saturation loading capacity. Both histograms have been truncated at 20 molecules per unit cell.

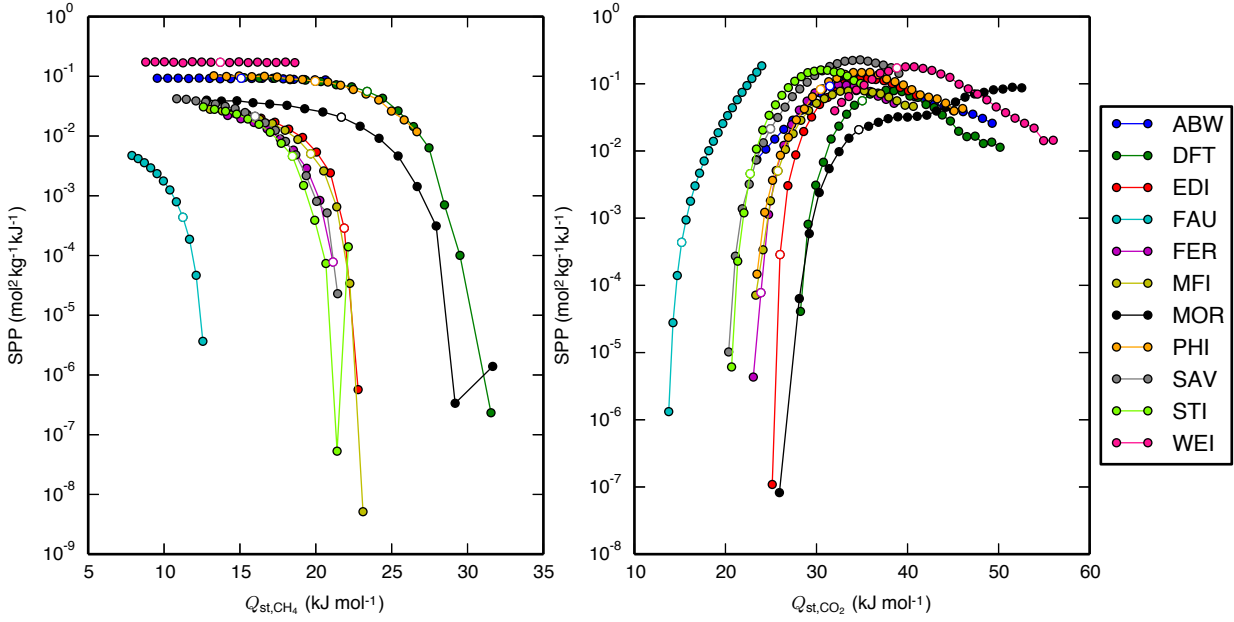


Fig. S11: SPP of a subset of IZA zeolites undergoing the LFG PSA process as a function of (left) Q_{st,CH_4} and (right) Q_{st,CO_2} . The points with white marker color represent the data with original guest-host epsilon parameters of the Lennard-Jones potential (115.00 K for CH_4-O_{zeo} , 50.20 K for $C_{CO_2}-O_{zeo}$, and 84.93 K for $O_{CO_2}-O_{zeo}$). (left) The CH_4-O_{zeo} epsilon value was varied between 75 K and 155 K in increments of 5 K (4.35% of the original value), with larger values resulting in a larger Q_{st,CH_4} . (right) The $C_{CO_2}-O_{zeo}$ epsilon value was varied between 32.128 K and 88.603 K in increments of 2.259 K while the $O_{CO_2}-O_{zeo}$ epsilon value was concurrently varied between 54.354 K and 149.904 K in increments of 3.822 K (both 4.50% of the original values), with larger values resulting in a larger Q_{st,CO_2} . When Q_{st,CH_4} was brought too high or Q_{st,CO_2} was brought too low, the material would become unfit for the separation, so these points are not shown. Note that SPP of the zeolite WEI does not change with Q_{st,CH_4} because CH_4 loading is negligibly low at all Q_{st,CH_4} values.

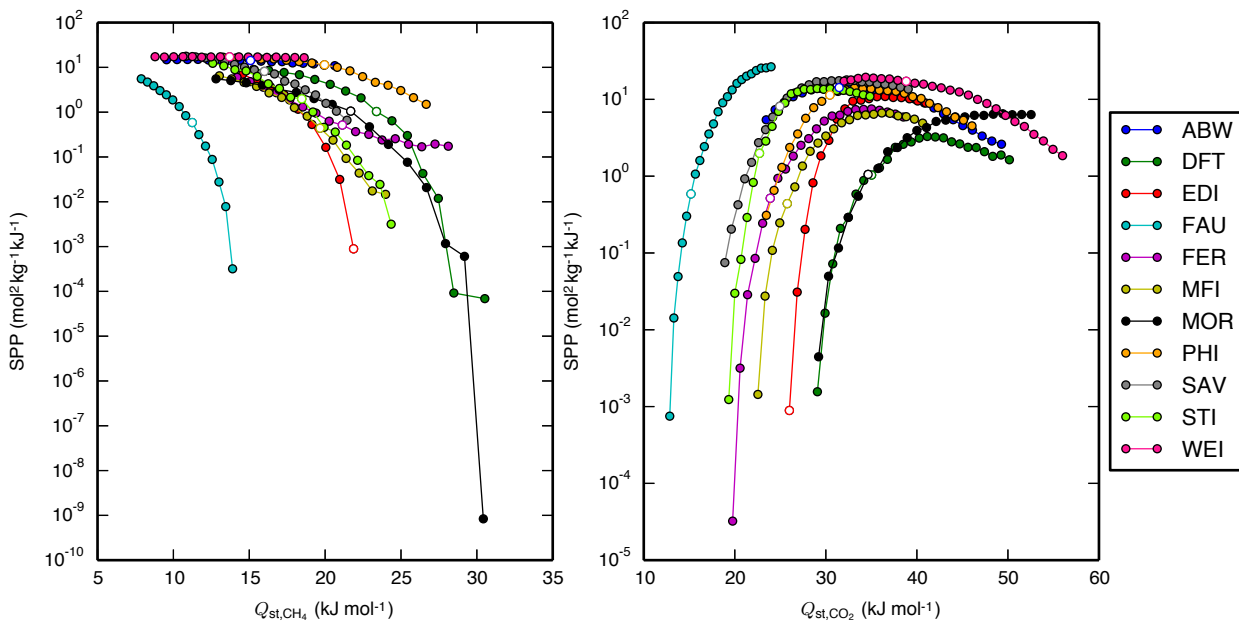


Fig. S12: SPP of a subset of IZA zeolites undergoing the NAG PSA process as a function of (left) Q_{st,CH_4} and (right) Q_{st,CO_2} . The points with white marker color represent the data with original guest-host epsilon parameters of the Lennard-Jones potential (115.00 K for CH_4-O_{zeo} , 50.20 K for $C_{CO_2}-O_{zeo}$, and 84.93 K for $O_{CO_2}-O_{zeo}$). (left) The CH_4-O_{zeo} epsilon value was varied between 75 K and 155 K in increments of 5 K (4.35% of the original value), with larger values resulting in a larger Q_{st,CH_4} . (right) The $C_{CO_2}-O_{zeo}$ epsilon value was varied between 32.128 K and 88.603 K in increments of 2.259 K while the $O_{CO_2}-O_{zeo}$ epsilon value was concurrently varied between 54.354 K and 149.904 K in increments of 3.822 K (both 4.50% of the original values), with larger values resulting in a larger Q_{st,CO_2} . When Q_{st,CH_4} was brought too high or Q_{st,CO_2} was brought too low, the material would become unfit for the separation, so these points are not shown. Note that SPP of the zeolite WEI does not change with Q_{st,CH_4} because CH_4 loading is negligibly low at all Q_{st,CH_4} values.

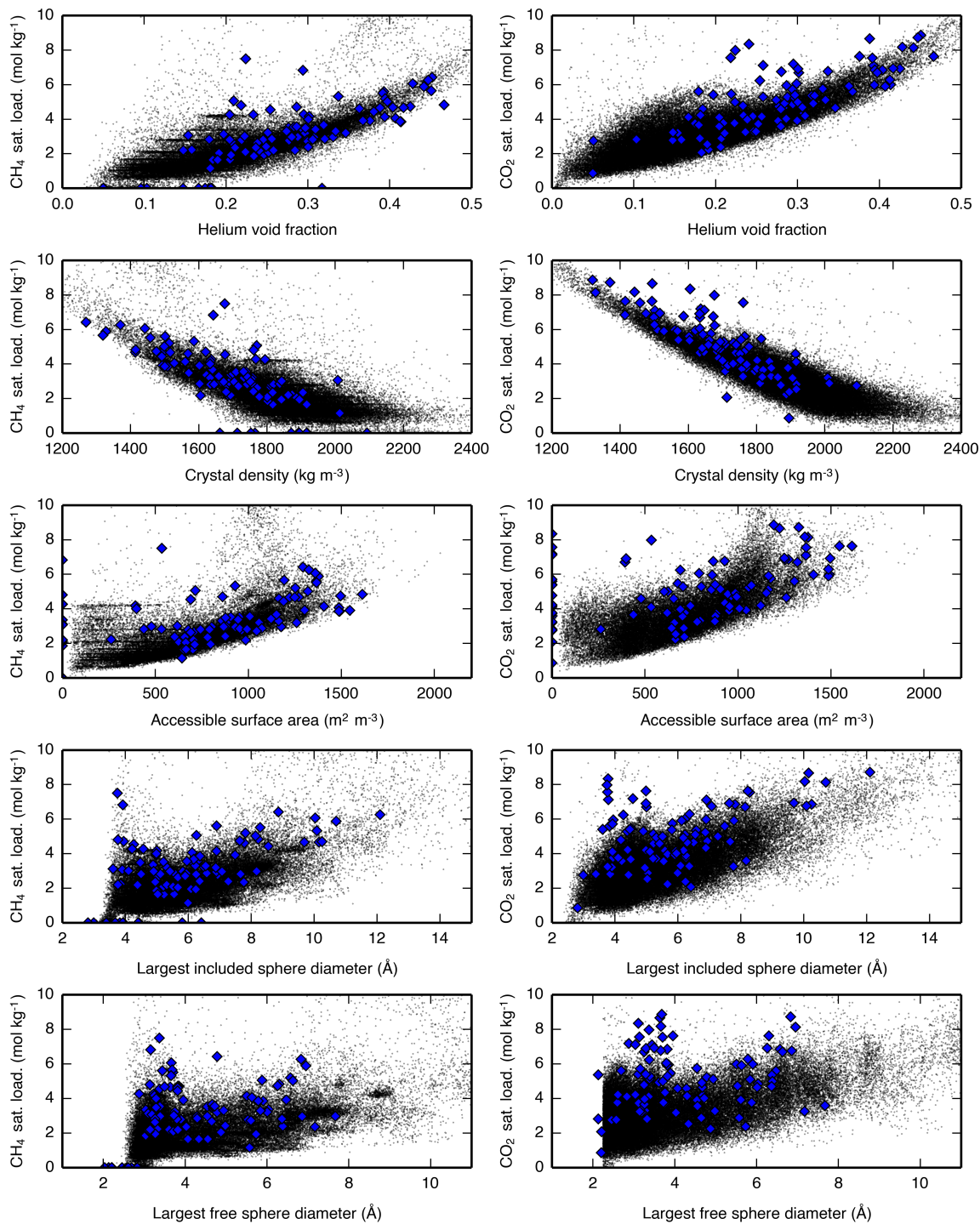


Fig. S13: Relationships between a zeolite's geometric descriptors and its saturation loading capacities. Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds. The stripes seen in plots containing CH_4 saturation loading capacity are due to that variable being more likely to be integer values of CH_4 molecules per unit cell (Fig. S10).

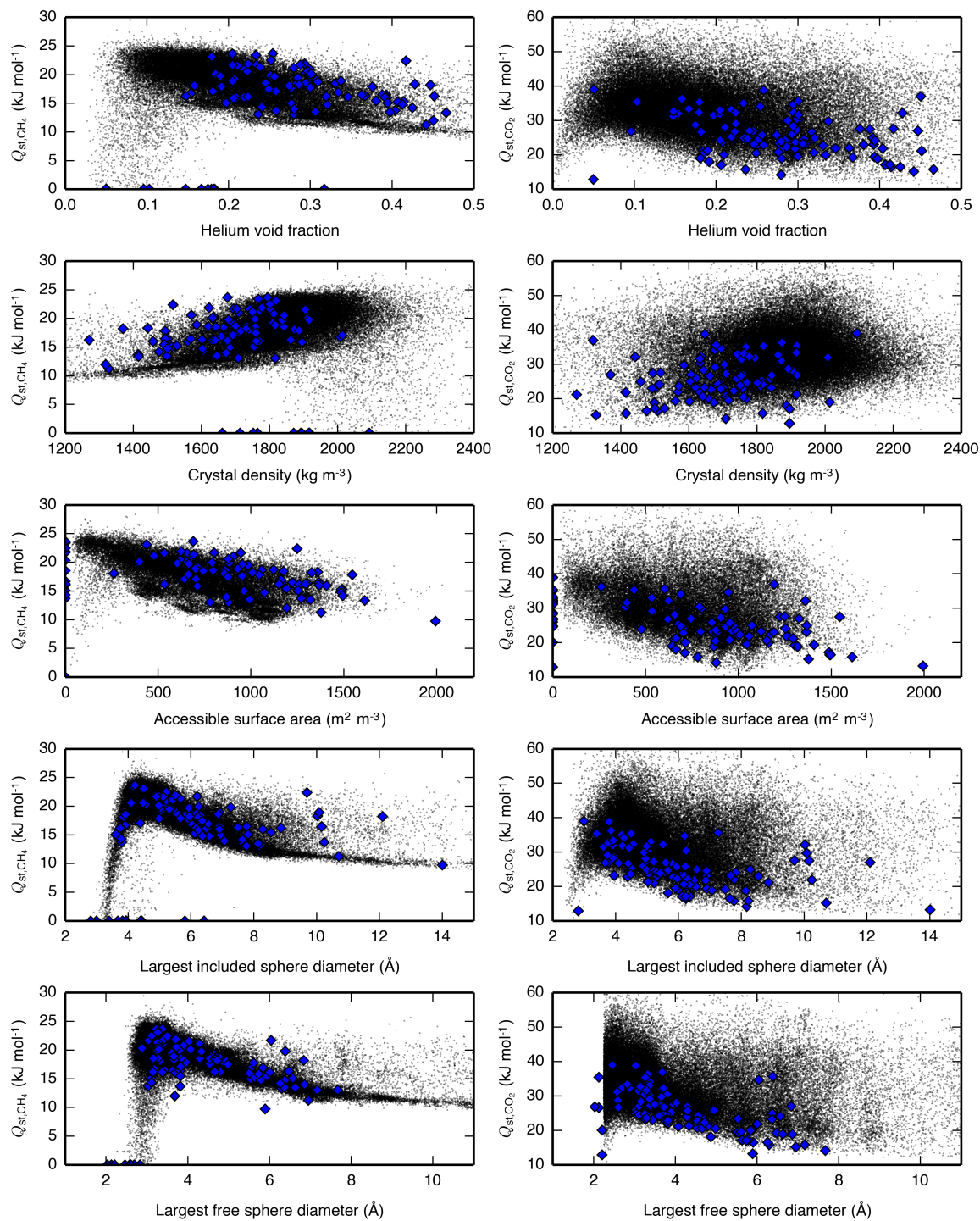


Fig. S14: Relationships between a zeolite's geometric descriptors and Q_{st} . Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds.

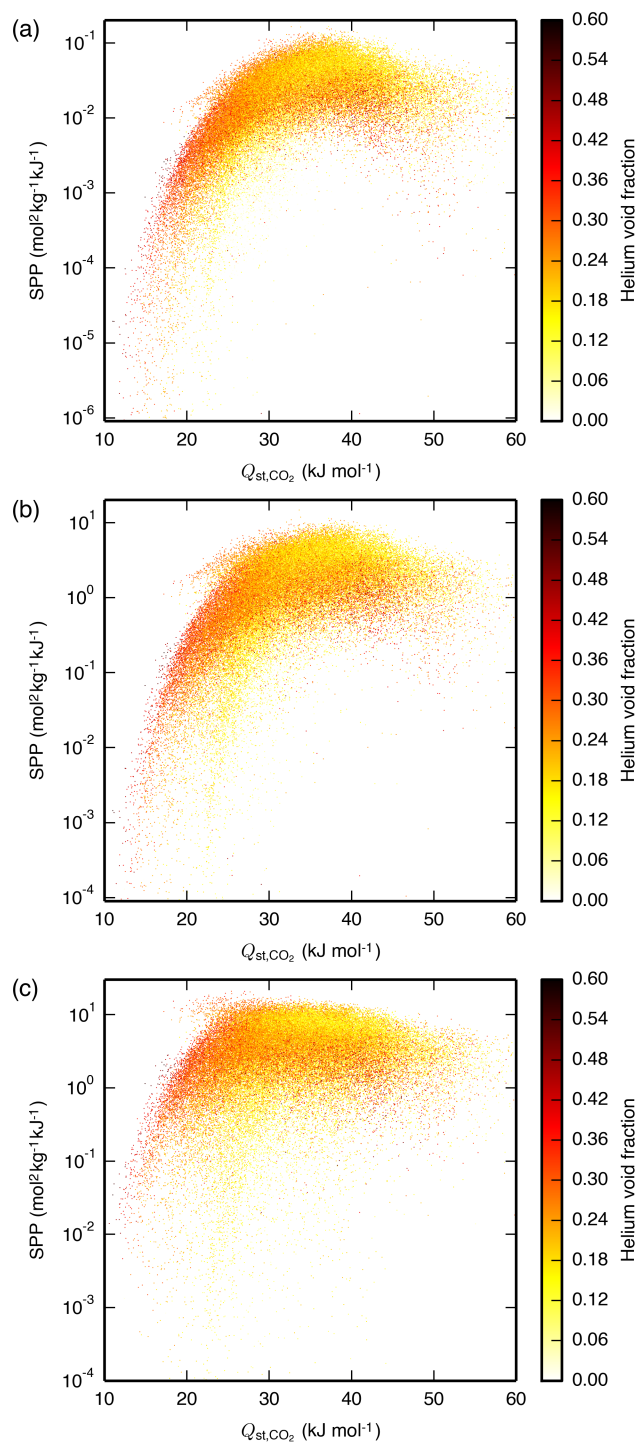


Fig. S15: SPP of the hypothetical zeolites as a function of Q_{st,CO_2} and helium void fraction, for the (a) LFG, (b) APG, and (c) NAG processes, all carried out with PSA. The materials are plotted in random order such that the data shown are representative of the materials hidden due to having similar Q_{st,CO_2} and SPP.

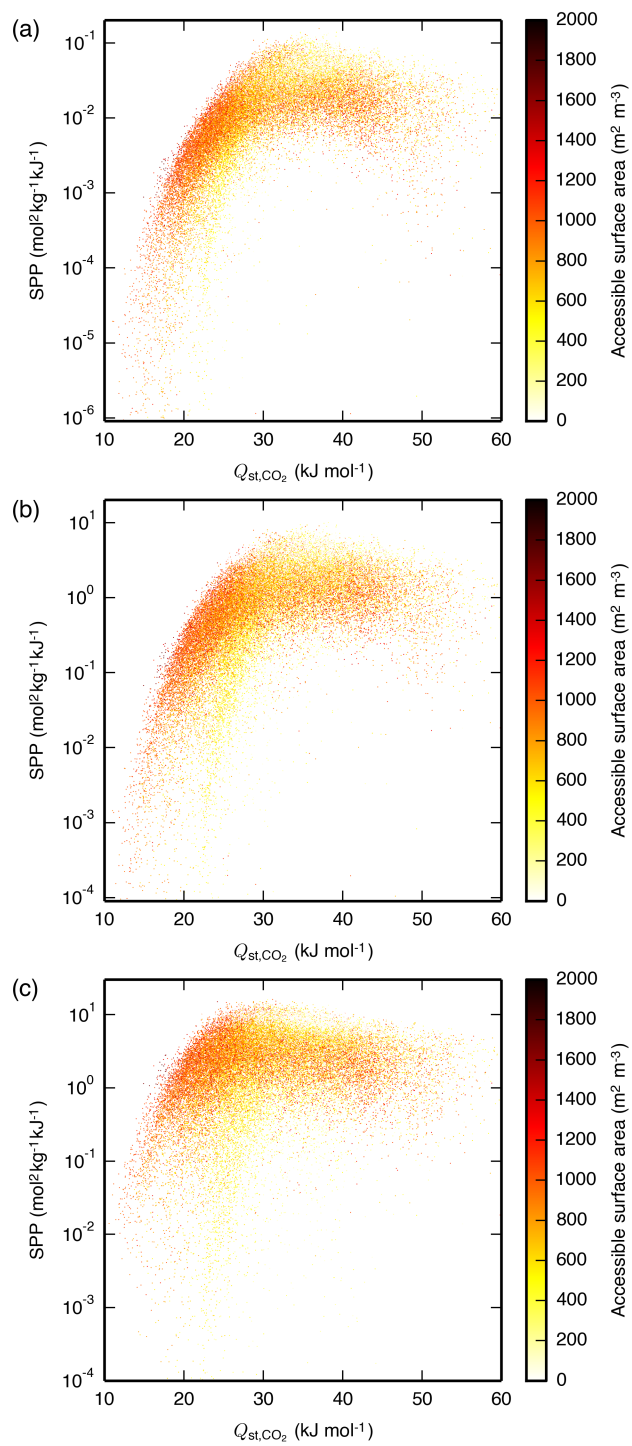


Fig. S16: SPP of the hypothetical zeolites as a function of Q_{st,CO_2} and accessible surface area, for the (a) LFG, (b) APG, and (c) NAG processes, all carried out with PSA. The materials are plotted in random order such that the data shown are representative of the materials hidden due to having similar Q_{st,CO_2} and SPP.

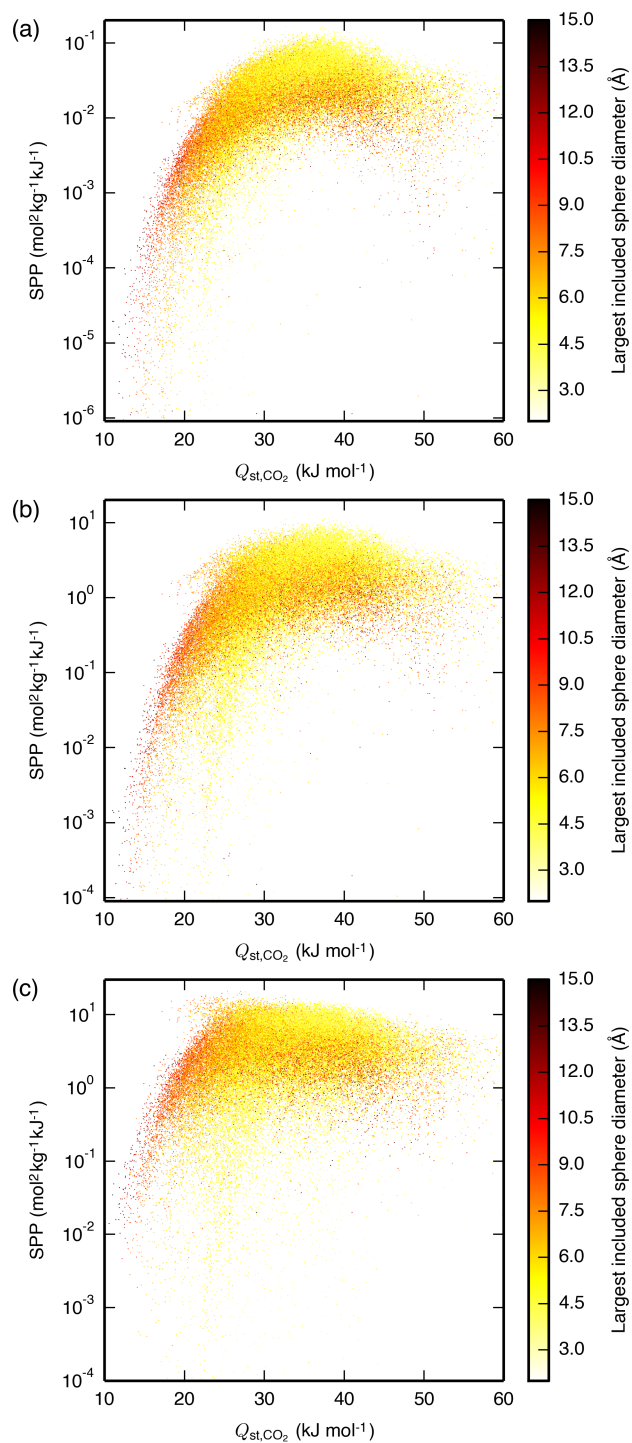


Fig. S17: SPP of the hypothetical zeolites as a function of Q_{st,CO_2} and largest included sphere diameter, for the (a) LFG, (b) APG, and (c) NAG processes, all carried out with PSA. The materials are plotted in random order such that the data shown are representative of the materials hidden due to having similar Q_{st,CO_2} and SPP.

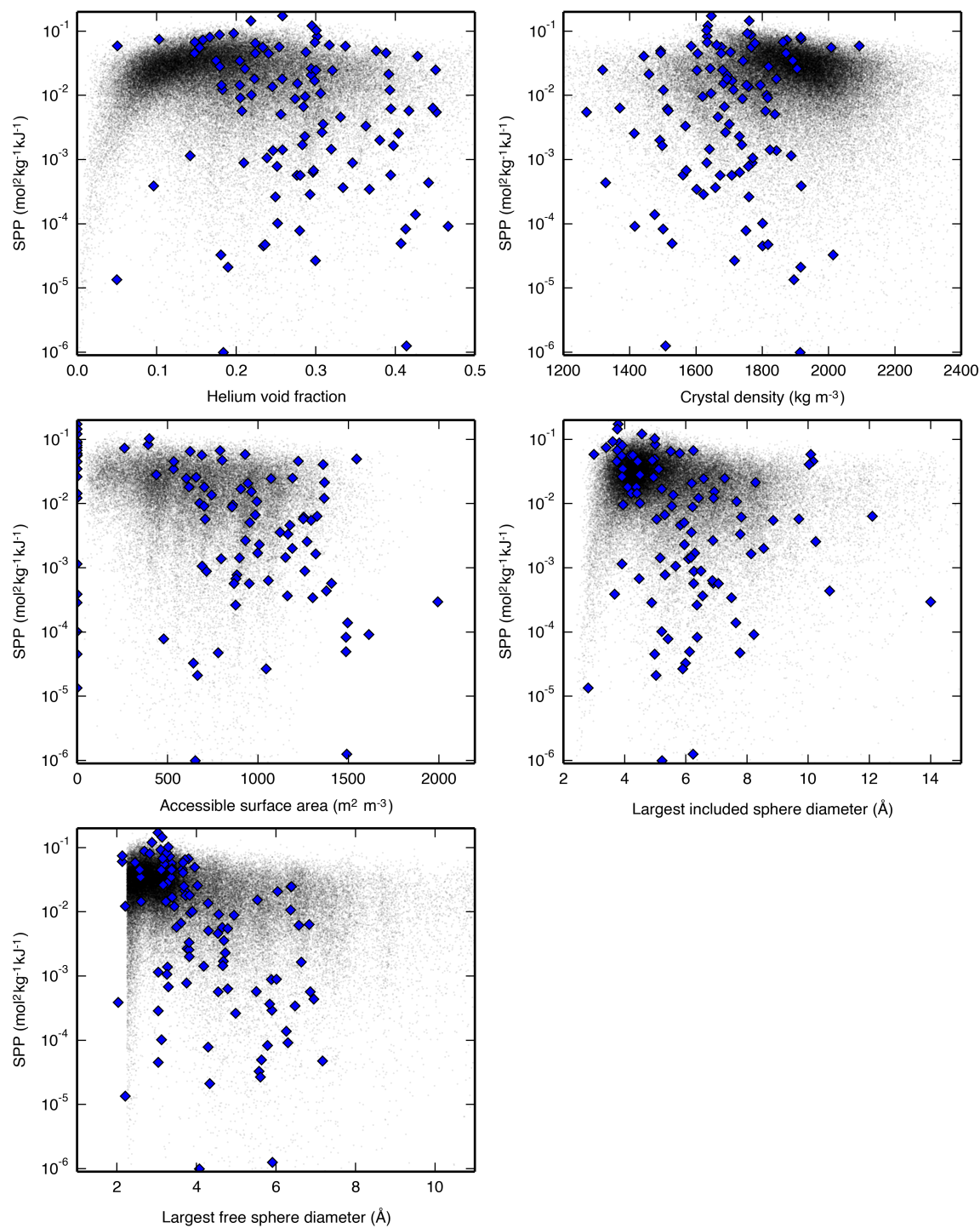


Fig. S18: SPP as a function of geometric parameters for the LFG process carried out with PSA. Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds.

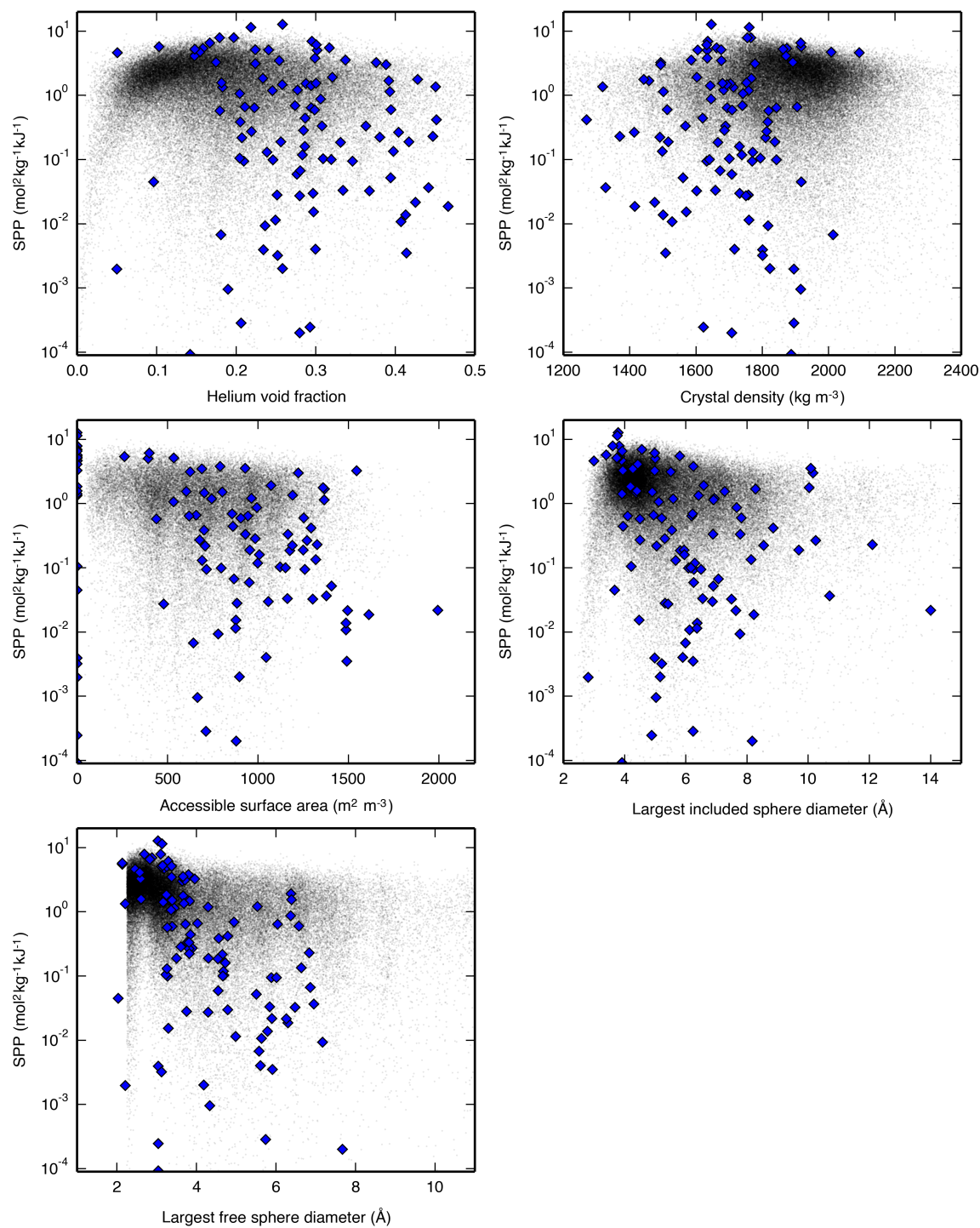


Fig. S19: SPP as a function of geometric parameters for the APG process carried out with PSA. Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds.

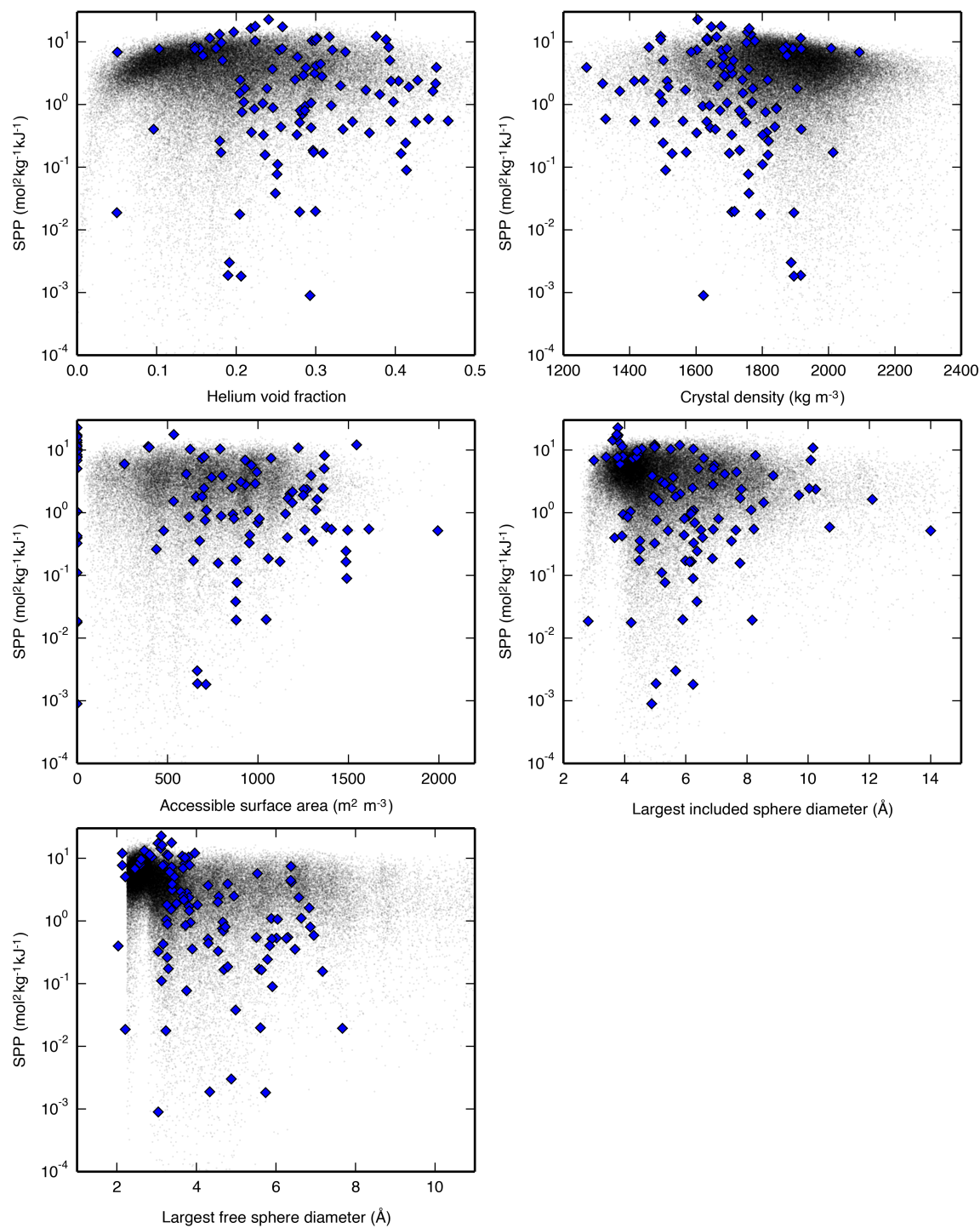


Fig. S20: SPP as a function of geometric parameters for the NAG process carried out with PSA. Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds.

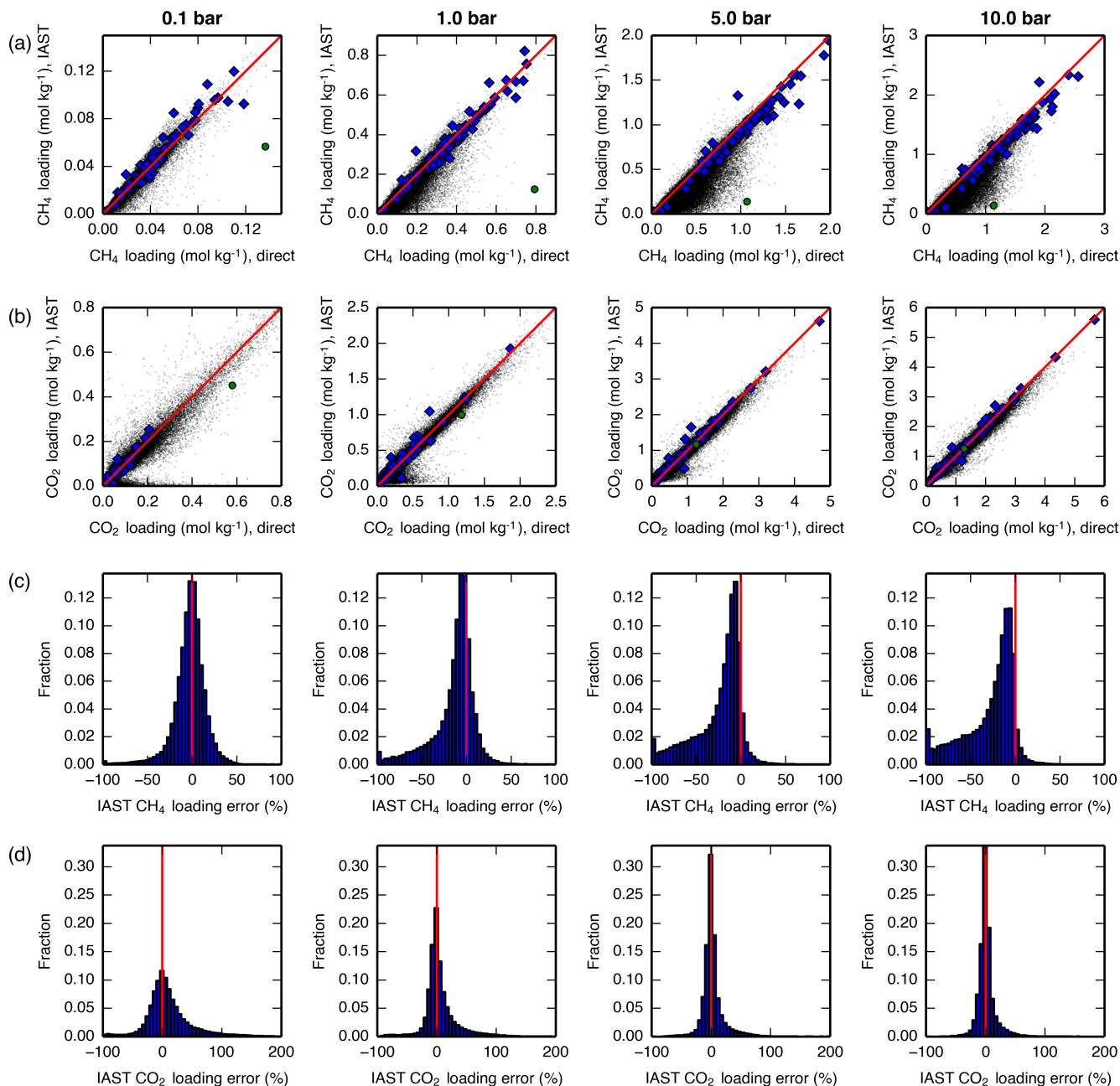


Fig. S21: Loadings of (a) CH_4 and (b) CO_2 at various total pressures, 300 K, and 90 mol % CH_4 . Data on the x-axis are taken from directly-simulated mixture isotherms, while data on the y-axis are taken from applying IAST to pure-component isotherms. A line is drawn at $y = x$ for reference. Hypothetical zeolites are shown as black dots, IZA zeolites are shown as blue diamonds, and hypothetical zeolite PCOD8205017 is shown as a green circle. Histograms of the fractional IAST error $\left(\frac{\text{IAST loading} - \text{direct loading}}{\text{direct loading}} \times 100\%\right)$ of the hypothetical zeolites only for (c) CH_4 and (d) CO_2 at the same conditions. A line is drawn at $x = 0$ for reference. All plots in the left-column are at 0.1 bar, all plots in the middle-left-column are at 1 bar, all plots in the middle-right-column are at 5 bar, and all plots in the right-column are at 10 bar. A similar plot showing results at 60 mol % CH_4 is given in Fig. 7.

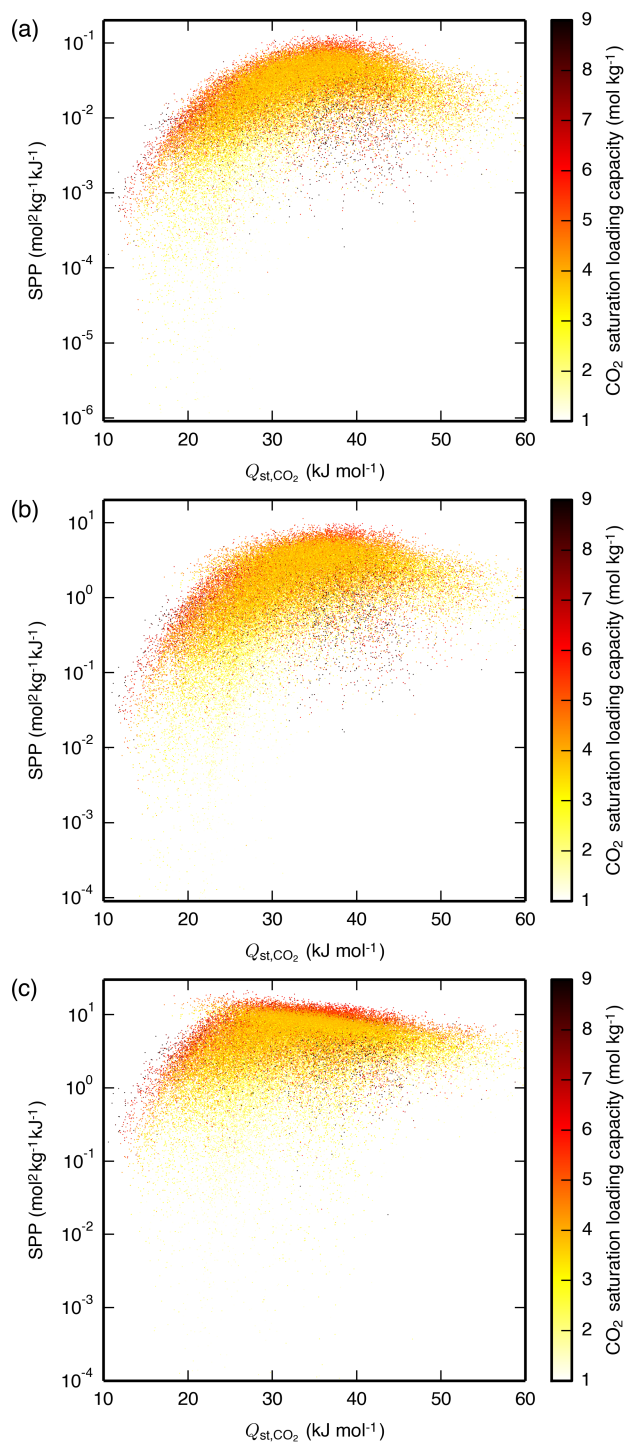


Fig. S22: SPP of the hypothetical zeolites as a function of Q_{st,CO_2} and the CO₂ saturation loading capacity for the (a) LFG, (b) APG, and (c) NAG processes, all carried out with PSA. The materials are plotted in random order such that the data shown are representative of the materials hidden due to having similar Q_{st,CO_2} and SPP. Mixture isotherms were obtained using IAST, whereas Fig. 2 used directly-generated mixture isotherms.

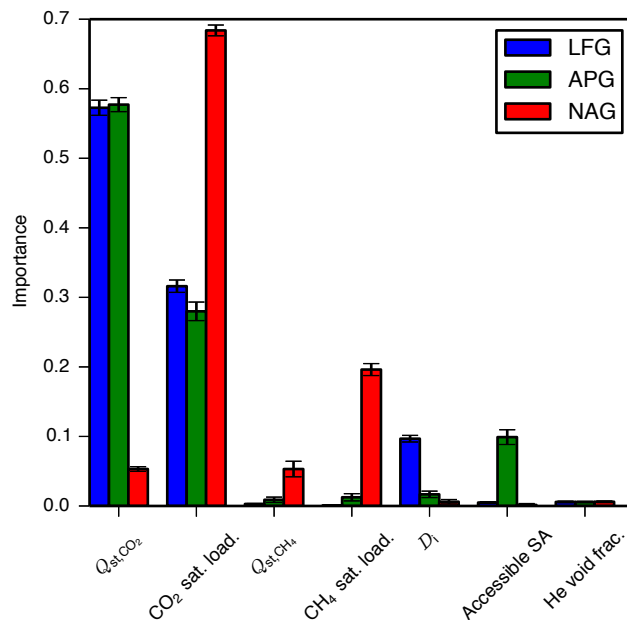


Fig. S23: Importance of geometric and isotherm descriptors to the SPP of the hypothetical zeolites undergoing PSA processes, as determined using a random forest of decision trees. The importance of a descriptor is calculated by summing the reductions in mean squared error brought about at each node where that descriptor splits a decision tree, averaging over all decision trees, and normalizing.¹⁹ Here, sat. load. is the saturation loading capacity, D_i is the largest included sphere diameter, Accessible SA is the accessible surface area, and He void frac. is the helium void fraction. Mixture isotherms were obtained using IAST, whereas Fig. 4 used directly-generated mixture isotherms.

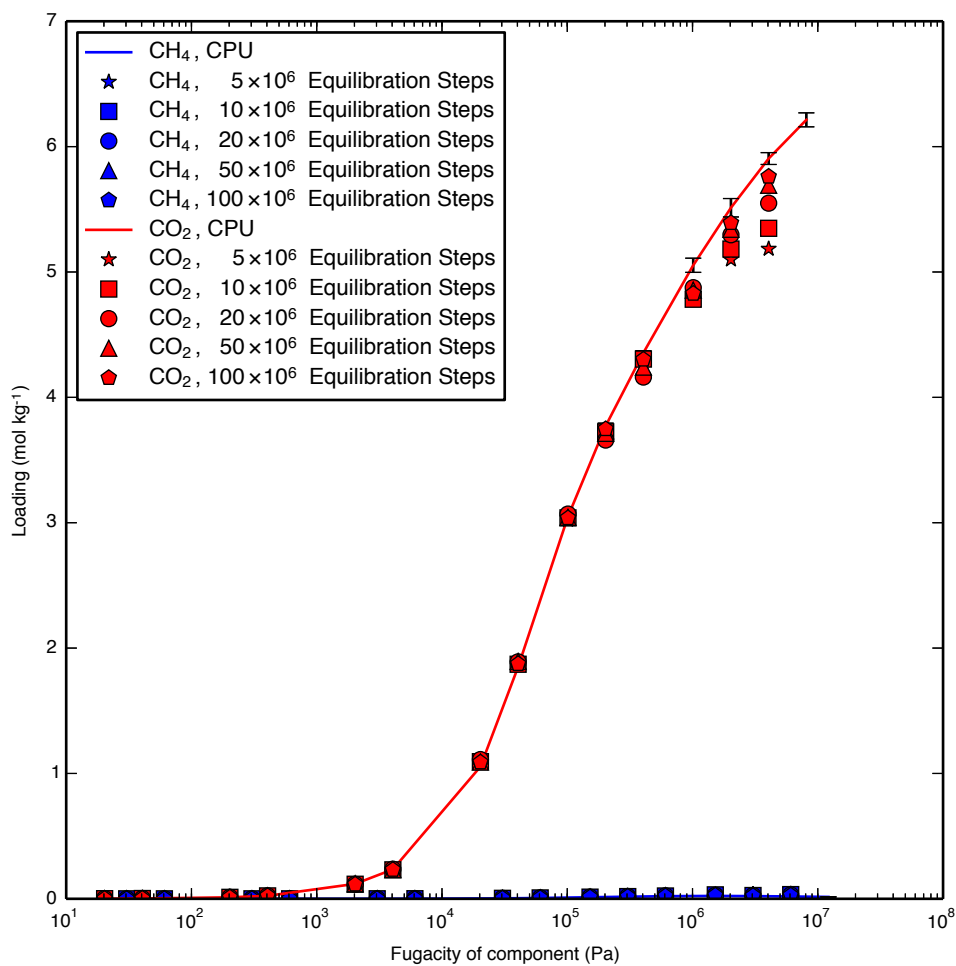


Fig. S24: Mixture isotherms of CH₄ (60 mol%) and CO₂ (40 mol%) in the IZA zeolite ABW, used to verify the GPU code. Unlike the isotherms used in the screening, these used the ideal gas equation of state and no pocket blocking. The CPU data were computed using 200,000 equilibration cycles and 100,000 production cycles, with error bars presented. The remainder of the data were generated using the GPU code with varying numbers of equilibration steps and 1,000,000 production steps.

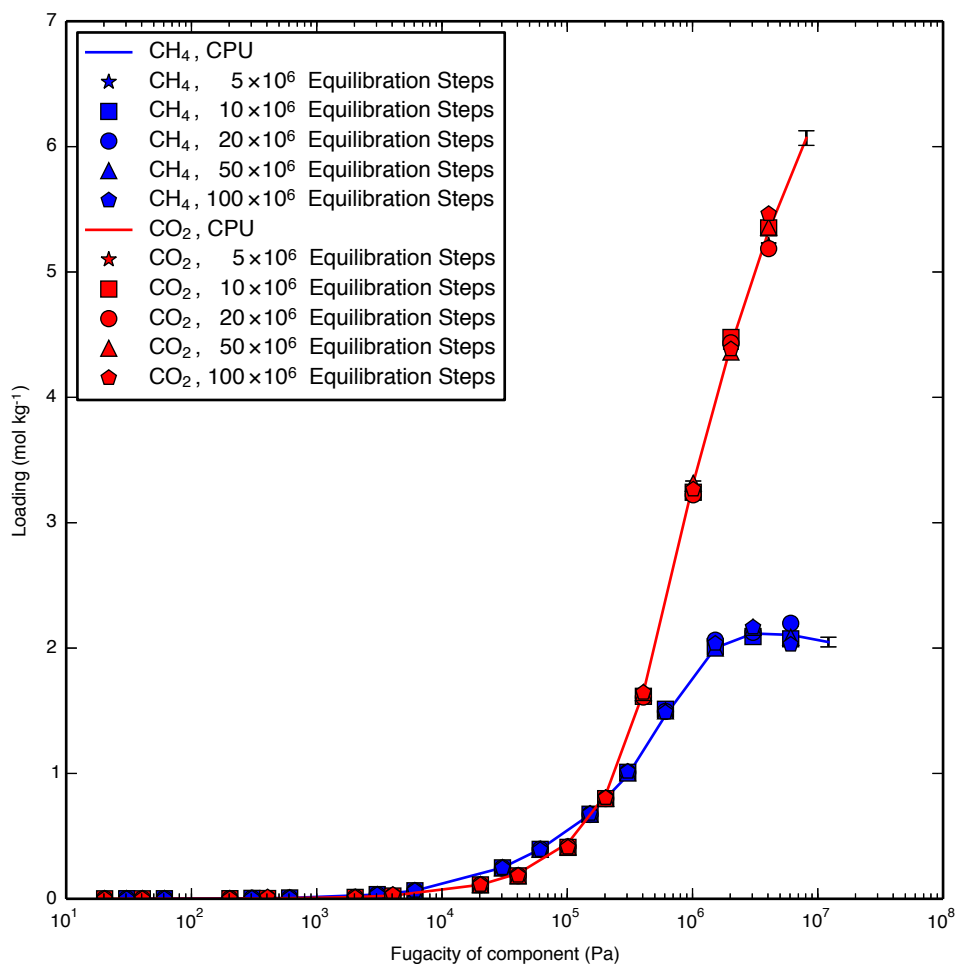


Fig. S25: Mixture isotherms of CH₄ (60 mol %) and CO₂ (40 mol %) in the IZA zeolite FAU, used to verify the GPU code. Unlike the isotherms used in the screening, these used the ideal gas equation of state and no pocket blocking. The CPU data were computed using 200,000 equilibration cycles and 100,000 production cycles, with error bars presented. The remainder of the data were generated using the GPU code with varying numbers of equilibration steps and 1,000,000 production steps.

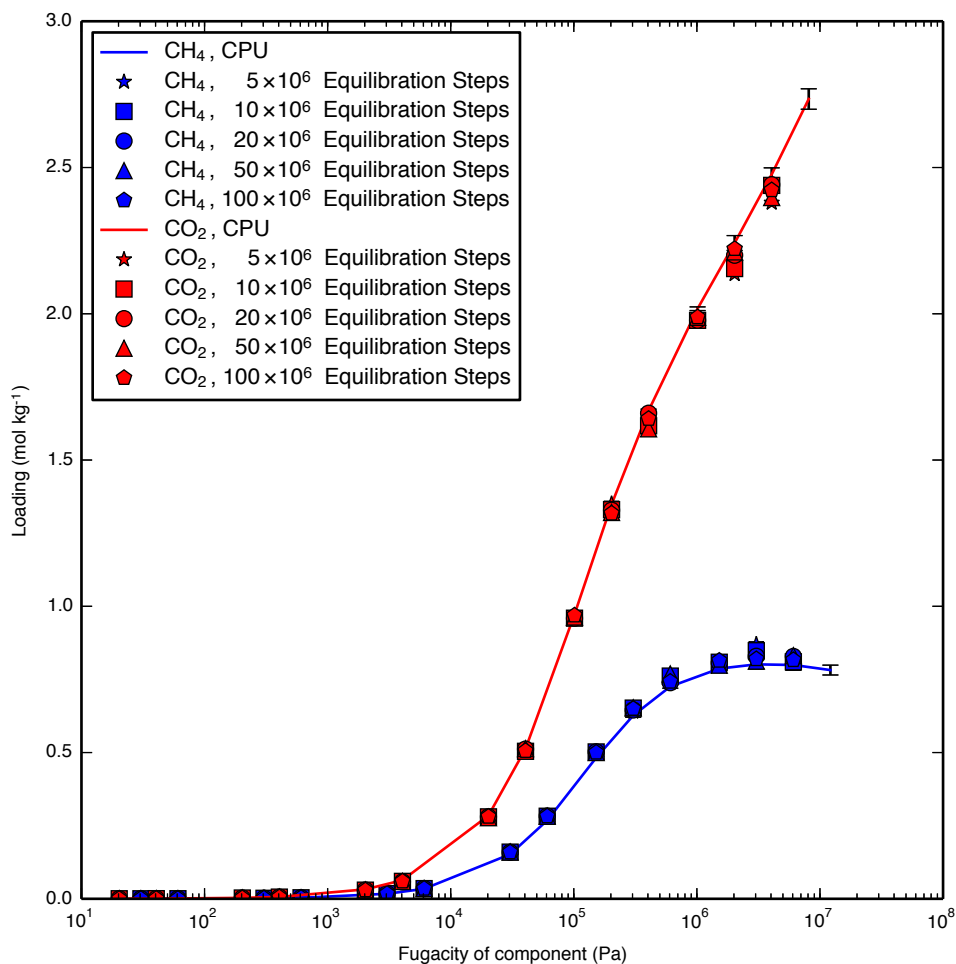


Fig. S26: Mixture isotherms of CH₄ (60 mol %) and CO₂ (40 mol %) in the IZA zeolite MFI, used to verify the GPU code. Unlike the isotherms used in the screening, these used the ideal gas equation of state and no pocket blocking. The CPU data were computed using 200,000 equilibration cycles and 100,000 production cycles, with error bars presented. The remainder of the data were generated using the GPU code with varying numbers of equilibration steps and 1,000,000 production steps.

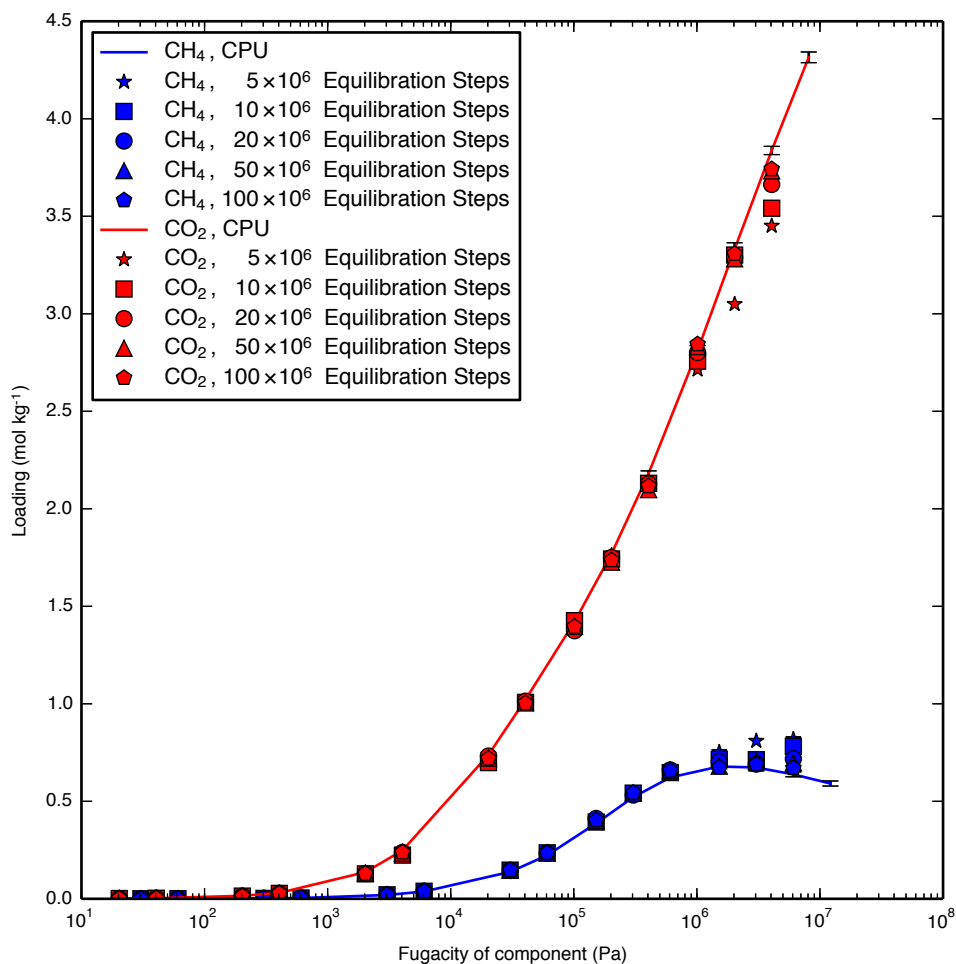


Fig. S27: Mixture isotherms of CH₄ (60 mol%) and CO₂ (40 mol%) in the IZA zeolite MOR, used to verify the GPU code. Unlike the isotherms used in the screening, these used the ideal gas equation of state and no pocket blocking. The CPU data were computed using 200,000 equilibration cycles and 100,000 production cycles, with error bars presented. The remainder of the data were generated using the GPU code with varying numbers of equilibration steps and 1,000,000 production steps.

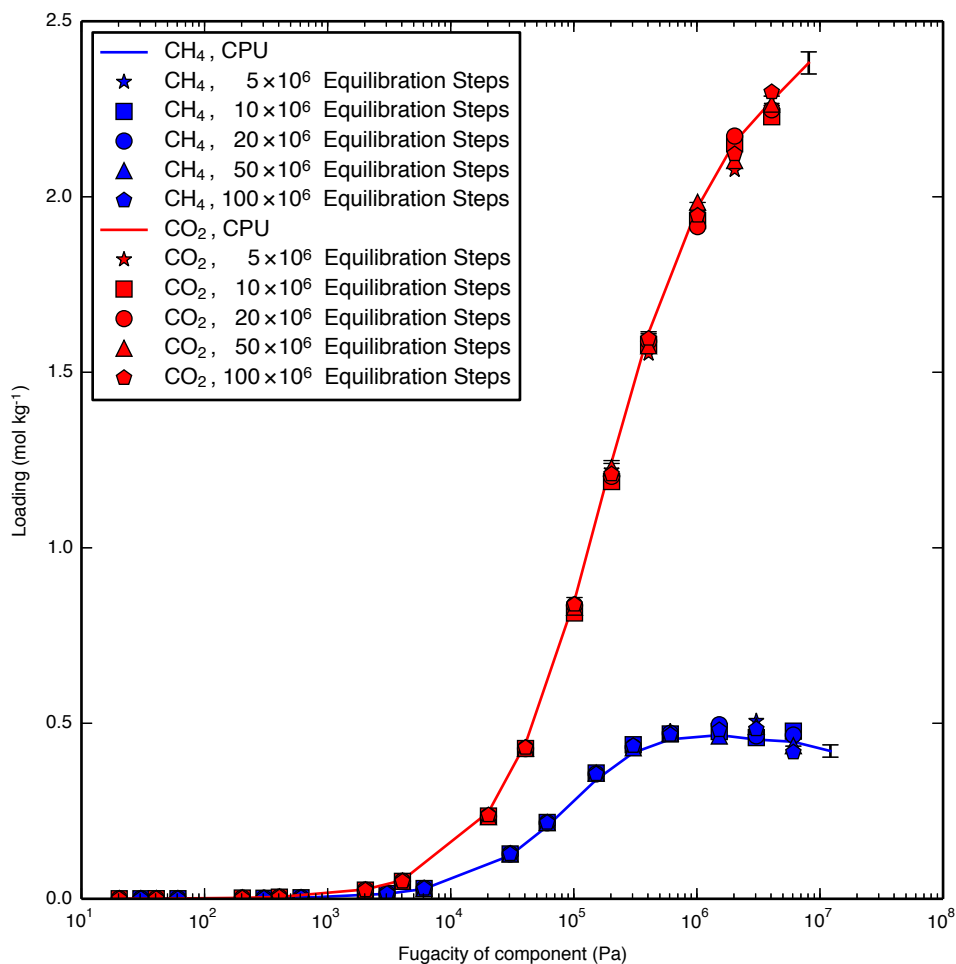


Fig. S28: Mixture isotherms of CH₄ (60 mol%) and CO₂ (40 mol%) in the IZA zeolite TON, used to verify the GPU code. Unlike the isotherms used in the screening, these used the ideal gas equation of state and no pocket blocking. The CPU data were computed using 200,000 equilibration cycles and 100,000 production cycles, with error bars presented. The remainder of the data were generated using the GPU code with varying numbers of equilibration steps and 1,000,000 production steps.

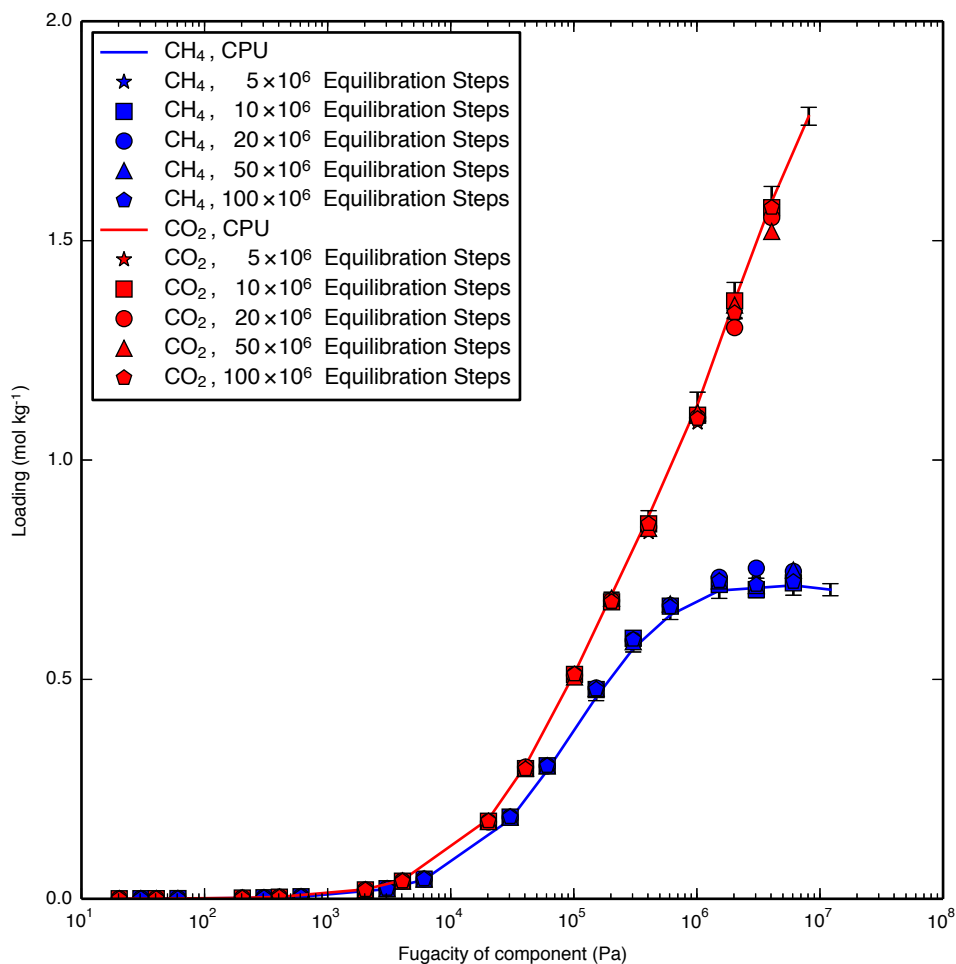


Fig. S29: Mixture isotherms of CH₄ (60 mol%) and CO₂ (40 mol%) in the hypothetical zeolite PCOD8170391, used to verify the GPU code. Unlike the isotherms used in the screening, these used the ideal gas equation of state and no pocket blocking. The CPU data were computed using 200,000 equilibration cycles and 100,000 production cycles, with error bars presented. The remainder of the data were generated using the GPU code with varying numbers of equilibration steps and 1,000,000 production steps.

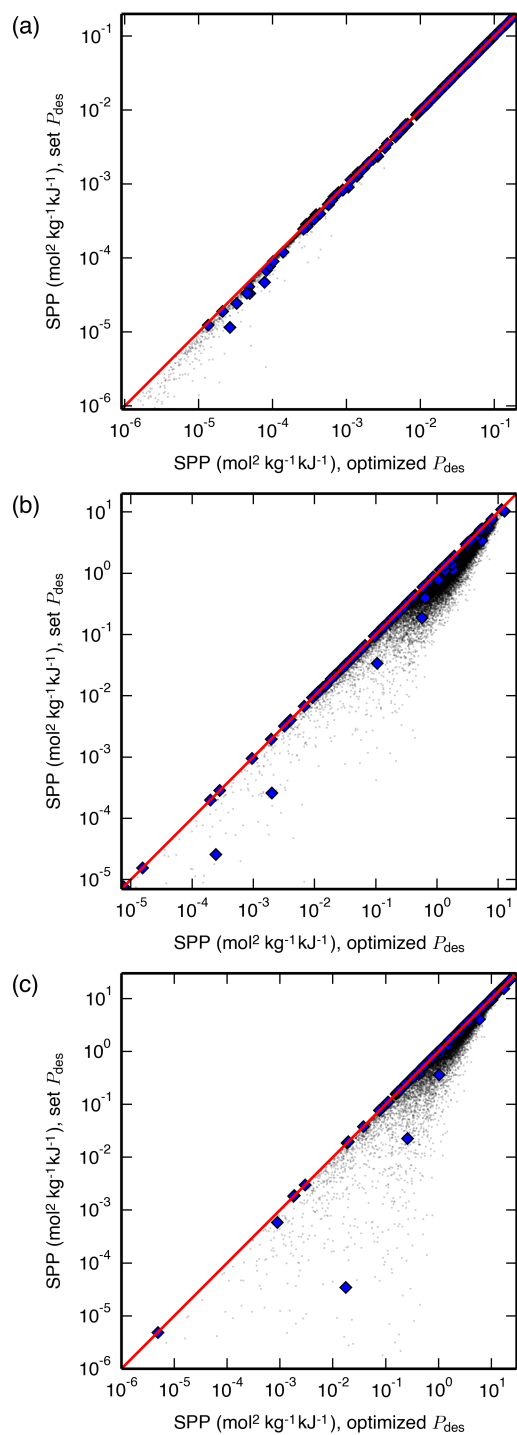


Fig. S30: A correlation between the SPP of materials evaluated at a set desorption pressure and the SPP evaluated at each material's optimal desorption pressure for the (a) LFG, (b) APG, and (c) NAG processes undergoing PSA. The LFG process used a set desorption pressure of 0.1 bar and the APG and NAG processes used a set desorption pressure of 1 bar, which might have been suspected as near-optimal desorption pressures for these processes prior to completing the present study.²⁰ Hypothetical zeolites are shown as black dots, and IZA zeolites are shown as blue diamonds. A line is drawn at $y = x$ for reference.

4 Supplementary Tables

All data in the following tables were taken from the analysis of PSA processes. Therefore all desorption temperatures are 300 K and all sensible energies are 0 kJ mol^{-1} .

Zeolites AST, LTN, MEP, NON, SGT, and UOZ were found to have a negative $Q_{\text{st},\text{CO}_2}$ and thus could not be used for any separations, so they are not included in the tables. In addition, some zeolites had a positive $Q_{\text{st},\text{CO}_2}$ but were found to have a required mass of adsorbent that was negative for a particular separation; this meant that they were unsuitable for that separation, and so they are not included in the tables. These include the zeolites GON, MRE, and SAF for the LFG PSA process, the zeolite MRE for the APG PSA process, and the zeolites ATV, BOF, and MRE for the NAG PSA process.

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