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## Motivation

### Goal

- Design optimal devices for adsorption-based Direct Air Capture (DAC) technologies within the Topology Optimization (TopOpt) framework that are suitable for scale-up.

### State-of-Art

- The use of DAC is imperative to handle hard-to-avoid CO<sub>2</sub> emissions [1].
- Contactors design improvements paired with sorbent material development is necessary for optimal devices that can be scaled-up.
- TopOpt has been used successfully for optimal structures in complex systems like flow batteries [2] and porous electrodes [3].
- Computational Fluid Dynamics (CFD) is the only tool that can quantitatively predict the effect of design parameterization on device performance.**
- Combining CFD and TopOpt is a powerful tool for seeking optimal device design for a given material.**

### Approach

- Topology Optimization is a mathematical framework used to determine the optimal material distribution within a given design domain with respect to a certain objective function (e.g. adsorption rate).

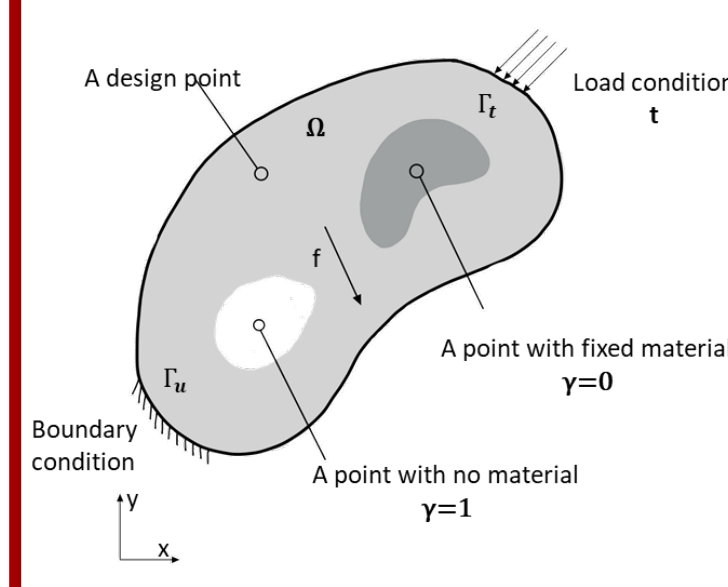
### Challenges

- Unique application of topology optimization.
- Considering multi-scale, packed bed adsorbers with complex, coupled physics within the Topology Optimization Framework.
- Defining representative nondimensional numbers that are appropriate for scaling up.
- Uniqueness of solution and global optimization not ensured a priori.

## Topology Optimization Framework

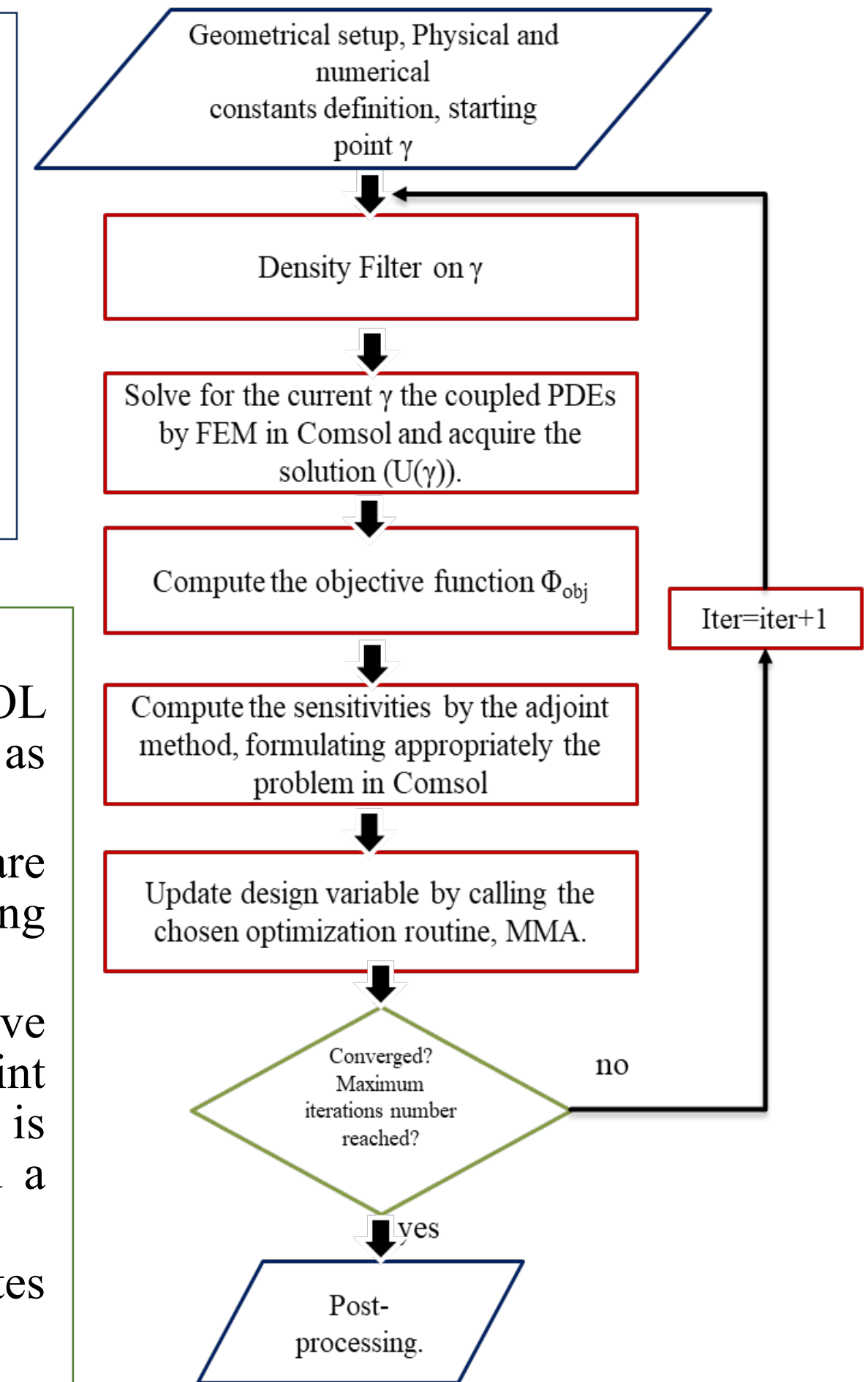
### Concept

- The computational domain  $\Omega$  with flow boundary conditions on  $\partial\Omega$  contains material distributed as solid and void, based on the local  $\gamma(\mathbf{r})$  values of each element. The local inverse permeability  $\alpha(\mathbf{r})$  in the N-S equations is parameterized using the local  $\gamma(\mathbf{r})$  [4], ranging between the limits of the open pore  $\gamma = 1$  and dense material  $\gamma = 0$ .



### Methodology

- The iterative algorithm is carried out in the software COMSOL Multiphysics through LiveLink™, the COMSOL/MATLAB interface, as follows:
  - The Navier Stokes and mass transport equations at the k<sup>th</sup> iteration are solved with a specific “guess” of the design variable  $\gamma^k(\mathbf{r})$  producing the finite element solution  $u^k(\gamma(\mathbf{r}))$ .
  - Sensitivity analysis is performed, where the gradient of the objective function and associated constraints are evaluated as well as the adjoint equations for the Lagrange multipliers  $\mu^k$ . The design variable is updated using the method of moving asymptotes (MMA) to yield a new guess  $\gamma^{k+1}$ .
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## Problem Statement

We wish to optimize one step in a post-combustion Pressure Swing Adsorption (PSA) process [5]. A feed containing N<sub>2</sub> and CO<sub>2</sub> is fed to the bed holding the sorbent.

The basis of this work is the formulation proposed by Olesen et al. [6]. The optimal design in [6] involves a complex  $\gamma$  distribution within the reactor. The algorithm optimizes the average reaction rate  $\Phi(\gamma) = -(k(\gamma)c_i)_{\Omega}$  by obtaining the optimal porosity field  $\gamma(\mathbf{r})$ , where  $-(k(\gamma)c_i)_{\Omega}$  is the kinetic reaction term. In our case, instead of reaction we treat adsorption instead for the DAC packed bed contactor. The overall constrained topology optimization model is:

$$\text{Max adsorption rate } \min_{\gamma} \Phi(\gamma) = -(k(\gamma)c_i)_{\gamma}$$

$$\text{s.t. } \int_{\Omega} \gamma(x) dx - |\Omega| \leq 0 \quad \text{Volume constraint}$$

$$\rho(\mathbf{u} \cdot \nabla) \mathbf{u} = -\nabla p + \eta \nabla^2 \mathbf{u} - \alpha(\gamma) \mathbf{u} \quad \text{N-S}$$

$$\nabla \cdot \mathbf{u} = 0 \quad \text{Continuity}$$

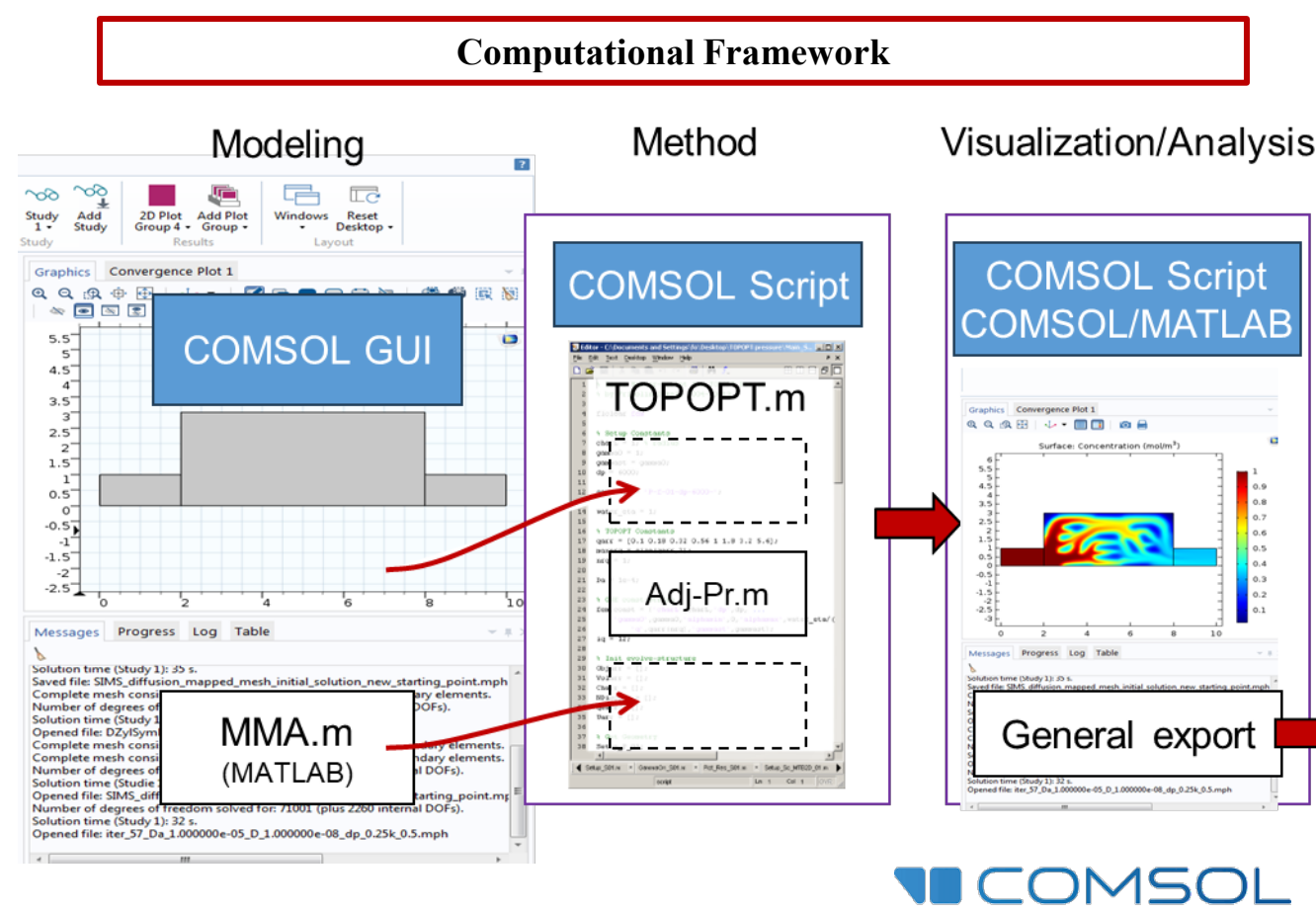
### Single or Dual-site Langmuir Isotherm

$$(1 - \epsilon_b) \rho_s k_i (q_i^* - q_i) = \frac{\partial (u c_i)}{\partial x}$$

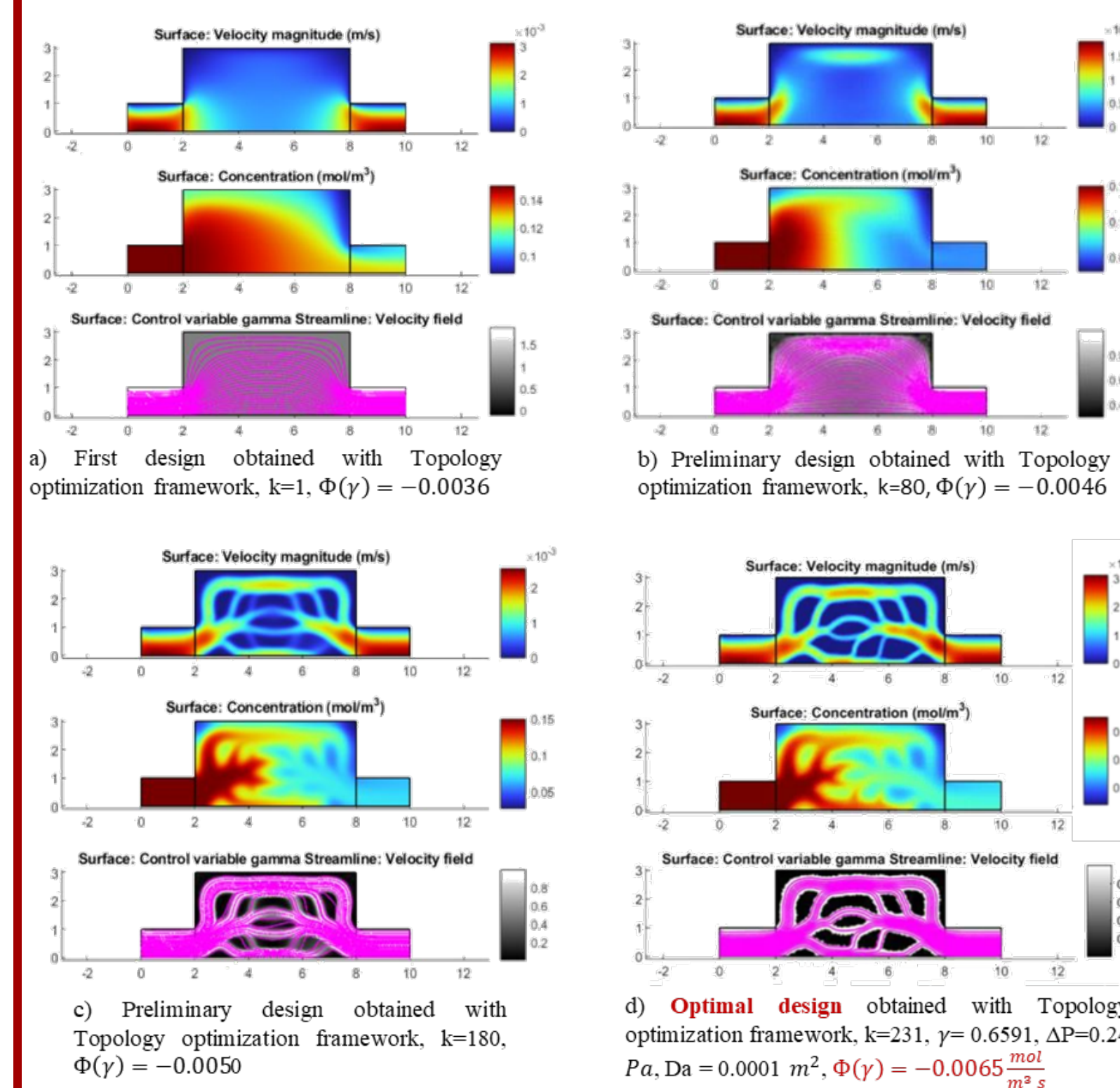
$$q_i^* = \frac{q_{i1} b_{i1} c_i RT}{1 + \sum_j b_{j1} c_j RT} + \frac{q_{i2} b_{i2} c_i RT}{1 + \sum_j b_{j2} c_j RT}$$

$$q_{mi}^s = k_{mi}^1 + k_{mi}^2 T, \quad b_{mi} = k_{mi}^3 \exp\left(\frac{k_{mi}^4}{T}\right), \quad m = 1, 2$$

$$0 \leq \gamma(\mathbf{r}) \leq 1 \quad \text{Design variable}$$



## Preliminary Results



### Model

- 3207 sec
- 231 iterations

### Size

- 3950 finite elements
- Triangular mesh
- 58011 degrees of freedom

- ❖ We obtained an optimal design for the process. The optimal value for the adsorption rate is 0.0065.
- ❖ **The acquired design increases the adsorption rate by 81% compared to the original design.**
- ❖ This value was achieved through the optimal material distribution that enhances the contact between the porous material and the gas stream.

## Conclusions

### Base

- Catalytic-packed bed reactors can be optimized to share underlying scaling properties. Therefore, this has the potential as a good starting point for packed bed DAC devices.

### Our model

- The adsorption rate of the packed bed can be significantly increased by distributing the active porous material within the device using topology optimization.

### Technology

- PSA processes could be a good starting point towards DAC design.
- Temperature Swing Adsorption (TSA) should be explored as well to determine the device potential.
- TSA process technology should be explored.

### Future work

- TSA/PSA with packed bed physics within the TopOpt framework.
- Determine and evaluate dimensionless numbers that characterize the system

## References

- [1] Liu, H., Ampah, J. D., Zhao, Y., Sun, X., Xu, L., Jiang, X., and Wang, S. (2022). A perspective on the overarching role of hydrogen, ammonia, and methanol carbon-neutral fuels towards net zero emission in the next three decades. *Energies*, (16).
- [2] Lin T.Y., Baker S.E., Duoss E.B. and Beck V.A. (2022). Topology Optimization of 3D Flow Fields for Flow Batteries. *Journal of The Electrochemical Society*. 169, 5.
- [3] Roy T., Salazar de Troya M.A., Worsley M.A., Beck V.A. (2022). Topology optimization for the design of porous electrodes. *Structural and Multidisciplinary Optimization* volume 65, Article number: 171.
- [4] Deng, Y., Wu, Y., and Liu, Z. (2018). Topology optimization theory for laminar flow
- [5] Agarwal, A., Biegler, L. T., and Zitney, S. E. (2009). A superstructure-based optimal synthesis of PSA cycles for post-combustion CO<sub>2</sub> capture. *AIChE Journal*, 56(7):1813–1828
- [6] Olesen, L. H., Okkels, F., and Bruus, H. (2006). A high-level programming-language implementation of topology optimization applied to steady-state navier-stokes flow. *International Journal for Numerical Methods in Engineering*, 65(7):975–1001.

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