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Effective Transport Properties



Pablo A. García-Salaberri

Abstract Porous media are an integral part of electrochemical energy conversion and storage devices, including fuel cells, electrolyzers, redox flow batteries and lithium-ion batteries, among others. The calculation of effective transport properties is required for designing more efficient components and for closing the formulation of macroscopic continuum models at the cell/stack level. In this chapter, OpenFOAM is used to determine the effective transport properties of virtually-generated fibrous gas diffusion layers. The analysis focuses on effective properties that rely on the fluid phase, diffusivity and permeability, which are determined by solving Laplace and Navier-Stokes equations at the pore scale, respectively. The model implementation (geometry generation, meshing, solver settings and postprocessing) is described, accompanied by a discussion of the main results. The dependence of orthotropic effective transport properties on porosity is examined and compared to traditional correlations ~~presented in the literature~~.

1 Introduction

Macroscopic continuum models are based on a volume-averaged formulation of mass, momentum, species, charge and energy conservation equations (Weber et al. 2014; Wang 2004; Goshtasbi et al. 2019; García-Salaberri et al. 2017). The model is closed through appropriate constitutive relationships that define the various effective properties of the cell components (García-Salaberri et al. 2018). Effective transport properties include the absolute permeability used in Darcy's law, the tortuosity factor used to correct Fick's law of diffusion, or the effective electrical and thermal

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1

22 conductivities used in Ohm's and Fourier's laws. However, these effective proper-
23 ties are challenging to determine in practice due to the thin nature of the porous
24 components used in electrochemical devices (thickness $\sim 10 - 1000 \mu\text{m}$), such as
25 gas diffusion layers and catalyst layers in fuel cells and active electrodes in batter-
26 ies (García-Salaberri et al. 2015a,b; Kashkooli et al. 2016; Liu et al. 2019). These
27 porous media must fulfill several critical functions, such as providing a transport
28 pathway for reactants/products through their pore volume and ensuring charge and
29 heat conduction through their solid structure. Catalyst layers and active electrodes
30 have the added functionality of providing a reactive surface area. Therefore, as a
31 complement to experimentation, numerical simulation at the pore scale has become
32 increasingly common. Pore-scale simulations in porous media provide direct insight
33 into the impact of the microstructure on transport processes, allowing one to deter-
34 mine effective transport properties and explore specific transport phenomena (see,
35 e.g., García-Salaberri et al. 2019; Hack et al. 2020; Sabharwal et al. 2016; Zhang et al.
36 2020; Gostick et al. 2007; Gostick 2013; Tranter et al. 2018; Belgacem et al. 2017;
37 Aghighi and Gostick 2017). A thorough understanding of the mass, charge and heat
38 transport properties of porous components is crucial for achieving improved perfor-
39 mance and durability.

40 Two main pore-scale modeling approaches are widely employed: pore-network
41 modeling (PNM) and direct numerical simulation (DNS) (Arvey et al. 2012). PNMs
42 idealize the pore space as a network of pore bodies interconnected by throats, whose
43 size and connectivity are determined from the microstructure of the porous media
44 (Gostick et al. 2007; García-Salaberri 2021). Some authors have also presented dual
45 networks that include both the solid phase and the standard fluid phase (Aghighi
46 and Gostick 2017). Different transport processes can be simulated on the net-
47 works, including capillary transport, convection, diffusion or heat conduction. In con-
48 trast, DNS solves the transport equations (e.g., Laplace or Navier-Stokes equations)
49 in computational meshes generated on tomography images or virtually-generated
50 microstructures of porous media. Numerical methods used to solve conservation
51 equations at the pore scale include the lattice Boltzmann method (LBM) or more
52 conventional techniques such as the finite-element (FEM) or finite-volume (FVM)
53 methods. Unlike the LBM, higher convergence rates are achieved with the FEM or
54 FVM using steady-state solvers, although the time invested in the mesh generation
55 step can represent a significant portion of the overall simulation (García-Salaberri
56 et al. 2015a,b; Liu et al. 2019; Sabharwal et al. 2016). DNS only requires the input
57 of the bulk properties of the constituents of the material (e.g., the bulk diffusion coef-
58 ficient for effective diffusivity or the kinematic viscosity for absolute permeability),
59 providing direct insight into the impact of microstructure on transport. Hence, the
60 information that can be potentially extracted from DNS is higher, but the computa-
61 tional cost is significantly higher compared to PNM.

62 Previous works that used OpenFOAM to simulate pore-scale transport phenomena
63 in porous components of electrochemical energy conversion and storage devices
64 are reviewed below. The literature survey includes works focused both on polymer
65 electrolyte fuel cells (PEFCs) and solid oxide fuel cells (SOFCs).

66 In terms of PEFCs, James et al. (2012) examined the effect of inhomogeneous
67 assembly compression on the effective electrical/thermal conductivity and diffusivity
68 in a commercial GDL (SGL SIGRACET 30BA). The microstructure was extracted by
69 means of X-ray computed tomography, then triangulated using the marching-cubes
70 algorithm, and finally converted into a volumetric mesh for simulation. The numerical
71 results showed that assembly compression significantly affects the effective transport
72 properties between the under-the-land and under-the-channel regions. In addition, a
73 notable decrease of the effective gas diffusivity was found compared to that predicted
74 by widely used correlations, such as Bruggeman correlation (Bruggeman 1935) and
75 the random fibre model of Tomadakis and Sotirchos (1993). Pharoah et al. (2011)
76 analyzed the effective electrical/ionic conductivity and gas diffusivity of catalyst
77 layers as a function of the volume fraction of carbon-Pt, ionomer and fluid phases.
78 The microstructure was virtually-generated using spherical particles. It was found
79 that Knudsen numbers in the pore space varied between the transition regime and
80 Knudsen regime, with higher pore radius leading to lower Knudsen number. In a
81 subsequent work, Khakaz-Babol et al. (2012) studied the coupling of transport and
82 electrochemical kinetics on microstructural representations of catalyst layers that
83 were generated using a similar algorithm to that of Pharoah et al. (2011). Different
84 Pt loadings were created by randomly exchanging Pt particles by carbon particles,
85 so that the base geometries were identical for each Pt loading. The results showed
86 that both the transport of protons and oxygen significantly affect performance, with
87 increased local losses in the ionomer at reduced Pt loadings.

88 Regarding SOFCs, Choi et al. (2009) analyzed the effective electrical and ionic
89 conductivity and gas diffusivity (including Knudsen diffusion) of the anode and
90 cathode electrodes. The microstructures were made of randomly distributed and
91 overlapping spheres with particle size distributions that matched those of ceramic
92 powders. The numerical results were compared against experimental data and the-
93 oretical correlations. Gunda et al. (2011) examined the effective transport prop-
94 erties (electrical conductivity and gas diffusivity) of ceramic lanthanum strontium
95 manganite (LSM) electrodes, whose ~~microstructure was~~ acquired using dual-beam
96 focused ion beam-scanning electron microscopy (FIB-SEM). The sensitivity of dif-
97 ferent image processing steps (threshold value, median filter radius, morphological
98 operators, surface triangulation, etc.) was examined. In addition, the work showed
99 that the effective transport properties ~~calculated~~ by FIB-SEM reconstruction were
100 more anisotropic than those ~~calculated~~ by numerical reconstruction. Next, Choi et
101 al (2011) presented a numerical framework for the computation of effective trans-
102 port properties of SOFC porous electrodes from ~~three-dimensional~~ reconstructions
103 of the microstructure based on measured parameters, such as porosity and particle
104 size distribution. Three different types of grids were considered: cartesian, octree,
105 and body-fitted/cut-cell with successive levels of surface refinement. OpenFOAM
106 was used to compute the effective transport properties in the three phases of the
107 electrode (pore, electron and ion). The model, validated with results from random
108 walk simulations, was used to investigate microstructures with monosized particle
109 distributions, as well as polydisperse particle size distributions similar to those found
110 in SOFC electrodes. Bertei et al. (2014) presented a modeling framework, based on

111 random sequential-addition packing algorithms, for the particle-based reconstruction
 112 of SOFC infiltrated electrodes. Key parameters, such as the connected triple-phase
 113 boundary length, effective electrical conductivity and effective diffusivity, were eval-
 114 uated on the reconstructed electrodes by using geometric analysis, FVM and random-
 115 walk methods. A parametric study showed that the critical loading (i.e., the perco-
 116 lation threshold) increases as the backbone porosity decreases and the nanoparticle
 117 diameter increases. Large triple-phase boundary length, specific surface area and
 118 good effective conductivity can be reached by infiltration, without detrimental effect
 119 on effective transport properties of the fluid phase.

120 In this chapter, ~~the reader is introduced to~~ DNS in porous media using Open-
 121 FOAM. Diffusion and convection are simulated in virtually-generated GDLs formed
 122 by a 2D arrangement of randomly-oriented fibres. The organization of the chapter
 123 is as follows. In Sect. 2, the physical model is presented, including the govern-
 124 ing equations, boundary conditions and calculation of effective transport properties,
 125 namely, effective diffusivity and permeability. In Sect. 3, the model implementation
 126 is described with a focus on geometry generation, meshing, solution procedure and
 127 post-processing. In Sect. 4, the results are discussed, including the effect of porosity
 128 on the computed orthotropic transport properties and a comparison with traditional
 129 correlations.

130 2 Physical Model

131 Diffusion and convection in the fluid phase of fibrous GDLs are examined. The gov-
 132 erning equations and the methodology used to determine the corresponding effective
 133 transport properties, effective diffusivity and absolute permeability, are presented
 134 below.

Effective Transport Properties of Solid Phase

Effective properties that rely on the solid phase (e.g., electrical conductivity) can be determined using a similar procedure to that presented in this chapter, but changing the phase of interest.

137 2.1 Diffusion: Effective Diffusivity

138 Species molar concentration C is determined from Laplace's equation

$$139 \quad \nabla \cdot (-D\nabla C) = 0 \quad (1)$$

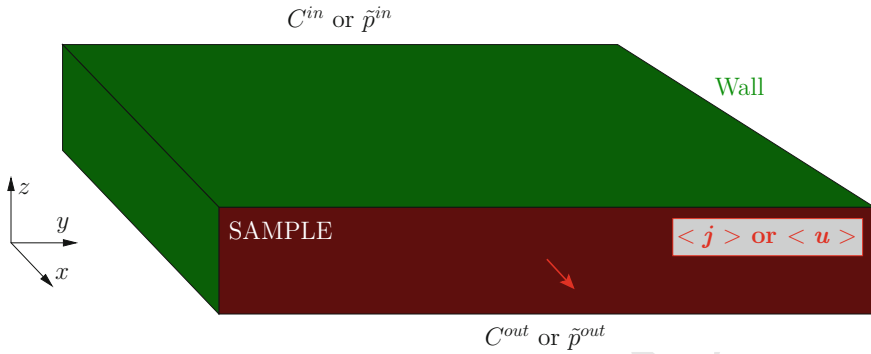


Fig. 1 Schematic of the computational domain and boundary conditions used to determine effective transport properties in the in-plane direction (x -direction). Similar boundary conditions are applied to compute effective transport properties in other directions. Effective diffusivity and permeability are determined from the computed average diffusive flux and velocity in the direction of interest, respectively

140 where D is the bulk diffusivity. This equation is subject to Dirichlet boundary conditions ($C = C^{\text{in}}$ and $C = C^{\text{out}}$) on the external faces of the domain perpendicular to the direction of interest i ($i = x, y$ or z) to create a concentration gradient. C^{in} and C^{out} are the inlet and outlet concentrations, respectively. A no-flux boundary condition ($\partial C / \partial \mathbf{n} = 0$) is set on the remaining faces of the domain and internal fluid-solid interfaces. A schematic representation of the external boundary conditions is shown in Fig. 1.

147 According to Fick's first law, the effective diffusivity of an anisotropic porous medium is given by a second-order tensor, whose diagonal and non-diagonal components can be determined by changing the direction of interest in the calculations

$$150 \quad \langle \mathbf{j} \rangle = -\bar{\bar{D}}^{\text{eff}} \nabla C; \quad \bar{\bar{D}}^{\text{eff}} = \begin{pmatrix} D_{xx}^{\text{eff}} & D_{xy}^{\text{eff}} & D_{xz}^{\text{eff}} \\ D_{yx}^{\text{eff}} & D_{yy}^{\text{eff}} & D_{yz}^{\text{eff}} \\ D_{zx}^{\text{eff}} & D_{zy}^{\text{eff}} & D_{zz}^{\text{eff}} \end{pmatrix} \quad (2)$$

151 where the symbol $\langle * \rangle$ denotes volume-average quantities. In this case, the average diffusive flux.

153 Since the flux vanishes in the solid region (s) of the porous medium, i.e., $\mathbf{j}_s = 0$, the diagonal components of the normalized effective diffusivity tensor (e.g., the zz -component) are given by

$$156 \quad \frac{D_{ii}^{\text{eff}}}{D} = \frac{\frac{1}{V_i} \int_{V_i} j_i dV}{\frac{D\Delta C}{L_i}} = \frac{\frac{V_f}{V_i} \left[\frac{1}{V_f} \int_{V_f} j_i dV \right]}{\frac{D\Delta C}{L_i}} = \varepsilon \frac{\langle j_i \rangle}{\frac{D\Delta C}{L_i}} \quad (3)$$

157 where V_i and V_f are the total and fluid (f) volume of the porous medium, respectively, L_i is the length of the domain in the direction of interest i , $\varepsilon = V_f / V_i$ is the porosity, and $j_i = -D \partial_i C$ is the diffusive flux in i -direction. Non-diagonal components were

160 not determined since they are typically small in the fibrous materials examined (see
161 the note below).

⚠ Non-Diagonal Components

The full effective diffusivity tensor $\bar{\mathbf{D}}^{\text{eff}}$ can be determined from three simulations (1,2,3) of diffusive flux fields (j_x^1, j_y^1, j_z^1) , (j_x^2, j_y^2, j_z^2) and (j_x^3, j_y^3, j_z^3) , corresponding to imposed concentration gradients in the x -, y - and z -directions, ∇C_x^1 , ∇C_y^2 and ∇C_z^3 , respectively.

Using Fick's first law, the components of the tensor are obtained by solving the following system of equations

$$-\begin{pmatrix} D_{xx}^{\text{eff}} & D_{xy}^{\text{eff}} & D_{xz}^{\text{eff}} \\ D_{yx}^{\text{eff}} & D_{yy}^{\text{eff}} & D_{yz}^{\text{eff}} \\ D_{zx}^{\text{eff}} & D_{zy}^{\text{eff}} & D_{zz}^{\text{eff}} \end{pmatrix} \begin{pmatrix} \nabla C_x^1 & \nabla C_x^2 & \nabla C_x^3 \\ \nabla C_y^1 & \nabla C_y^2 & \nabla C_y^3 \\ \nabla C_z^1 & \nabla C_z^2 & \nabla C_z^3 \end{pmatrix} = \begin{pmatrix} \langle j_x^1 \rangle & \langle j_x^2 \rangle & \langle j_x^3 \rangle \\ \langle j_y^1 \rangle & \langle j_y^2 \rangle & \langle j_y^3 \rangle \\ \langle j_z^1 \rangle & \langle j_z^2 \rangle & \langle j_z^3 \rangle \end{pmatrix} \quad (4)$$

where ∇C_x^2 , ∇C_x^3 , ∇C_y^1 , ∇C_y^3 , ∇C_z^1 and ∇C_z^2 are the average concentration gradients computed in the transverse directions.

Although in the simulations we prescribe a local no-flux boundary condition at the sidewalls of the domain, the average concentration gradients and average diffusive fluxes in the transverse directions are in general different from zero. Similar considerations apply for the permeability tensor [Guibert et al. \(2016\)](#).

162

2.2 Convection: Permeability

163

164 Convection is modeled through the steady-state mass conservation and Navier-Stokes
165 equations for an incompressible Newtonian fluid

$$\begin{aligned} \nabla \cdot \mathbf{u} &= 0 \\ \nabla \cdot (\mathbf{u}\mathbf{u}) &= -\nabla \tilde{p} + \nabla \cdot (\nu (\nabla \mathbf{u} + \nabla \mathbf{u}^T)) \end{aligned} \quad (5)$$

166

167 where \mathbf{u} is the velocity vector, ν is the kinematic viscosity, p is the static pressure and
168 $\tilde{p} = p/\rho_*$ with ρ the density. Similar to diffusion, Dirichlet boundary conditions are
169 prescribed for pressure on the external faces of the domain in the direction of interest
170 \hat{i} , $\tilde{p} = \tilde{p}^{\text{in}}$ and $\tilde{p} = \tilde{p}^{\text{out}}$. An impermeable no-slip boundary condition ($\mathbf{u} = 0$) is set
171 on the remaining faces of the domain and interior fluid-solid interfaces of the porous
172 medium.

173 According to Darcy's law,

$$174 \quad \langle \mathbf{u} \rangle = -\frac{\overline{\mathbf{K}}}{\nu} \nabla \tilde{p}; \quad \overline{\mathbf{K}} = \begin{pmatrix} K_{xx} & K_{xy} & K_{xz} \\ K_{yx} & K_{yy} & K_{yz} \\ K_{zx} & K_{zy} & K_{zz} \end{pmatrix}, \quad (6)$$

175 the diagonal components of the permeability tensor, K_{ii} , are determined as

$$176 \quad K_{ii} = \frac{\nu \left[\frac{1}{V_i} \int_{V_i} u_i dV \right]}{\frac{\Delta \tilde{p}}{L_i}} = \frac{\nu \frac{V_f}{V_i} \left[\frac{1}{V_f} \int_{V_f} u_i dV \right]}{\frac{\Delta \tilde{p}}{L_i}} = \varepsilon \nu L_i \frac{\langle u_i \rangle_f}{\Delta \tilde{p}} \quad (7)$$

177 where $\Delta \tilde{p} = \tilde{p}^{\text{in}} - \tilde{p}^{\text{out}}$ is the prescribed 'pressure' difference and u_i is the
178 i -component of the velocity vector.

179 Calculations of absolute permeability were performed in physical units, consid-
180 ering $\Delta \tilde{p} = 1 \text{ m}^2/\text{s}^2$ ($\tilde{p}^{\text{in}} = 1 \text{ m}^2/\text{s}^2$, $\tilde{p}^{\text{out}} = 0$), while ν was adjusted to ensure that
181 the flow was in the creeping regime, i.e.,

$$182 \quad Re = \frac{\langle \|\mathbf{u}\| \rangle d_f}{\nu} \approx \frac{\langle u_i \rangle d_f}{\nu} \ll 1 \quad (8)$$

183 where $\langle \|\mathbf{u}\| \rangle$ is the average modulus of the velocity and d_f is the diameter of the
184 mono-sized fibres. Using these values, the permeability is given by

$$185 \quad K_{i,i} = \varepsilon \nu L_i \langle u_{i,f} \rangle \quad (9)$$



186 where $\langle u_{i,f} \rangle$ is the volume-average velocity in i -direction within the fluid phase
187 computed in the simulations.

Darcy's Law

Darcy's law can be verified by varying $\Delta \tilde{p}$, while keeping the value of ν constant. The linear relationship between $\Delta \tilde{p}/L_i$ and Re shows that inertia is not important and the flow is indeed in the creeping regime, $Re \ll 1$.

190 3 Model Implementation

191 The main steps followed for the calculation of the effective transport properties
192 are presented in this section, including the geometry generation, meshing routine,
193 solver selection and post-processing. An example of the bash script used to run
194 the simulations of effective diffusivity (and permeability) is presented below. The
195 number of processors in the parallel execution is given as an argument to the script.
196 Before running the script, the user must generate the triangulated geometry of the

197 porous medium (*facets.stl*) using the *gdl.cpp* code. In addition to the solution fields,
 198 the output results include the average diffusive fluxes (or velocity component) in
 199 each direction, which are saved periodically (as indicated in *controlDict*) and in
 200 the last iteration. The computed values are written into the *postProcessing* folder.
 201 The average quantities corresponding to the  iteration are used to determine the
 202 effective properties through Eqs. (3) and (9).  Since the dimensions of the domain,
 203 porosity and bulk properties are known in advance.

Listing 3.1 Main script used to run simulations starting from the geometry file *facets.stl*.

```

204 #Argument: number of processors (equal to the number of
205     subdomains in decomposeParDict)
206
207 #Clean folders
208 foamListTimes -rm
209 rm -r ./postProcessing
210 rm -r ./processor*
211
212 #####
213 #MESHING
214 #####
215
216 #Run blockMesh
217 blockMesh > log
218
219 #Copy files
220 cp facets.stl ./constant/triSurface
221
222 #Run snappyHexMesh
223 decomposePar >> log
224 mpirun -np $1 snappyHexMesh -overwrite -parallel >> log
225
226 #Set initial fields
227 ls -d processor* | xargs -I {} rm -rf ./{}/*
228 ls -d processor* | xargs -I {} cp -r 0.orig ./{}/*
229
230 #Scale mesh to meters [m]
231 mpirun -np $1 transformPoints -scale "(0.001 0.001 0.001)" -
232     parallel >> log
233
234 #Renumber mesh
235 mpirun -np $1 renumberMesh -overwrite -parallel >> log
236
237 #Check final mesh
238 mpirun -np $1 checkMesh -parallel >> log
239
240 #####
241 #RUN SOLVER
242 #####
243

```

```

244 (diffusion) mpirun -np $1 laplacianFoam -parallel -
245     noFunctionObjects >> log
246 (convection) mpirun -np $1 simpleFoam -parallel -
247     noFunctionObjects >> log
248
249 #####
250 #POSTPROCESSING
251 #####
252
253 #Create cell region
254 mpirun -np $1 topoSet -parallel >> log
255
256 #Calculate average fluxes/velocities
257 (diffusion) mpirun -np $1 postProcess -fields "(T gradTx gradTy
258     gradTz)" -parallel >> log
259 (convection) mpirun -np $1 postProcess -fields "(U p)" -
260 parallel >> log
261

```

Managing Computational Simulations

Since the meshing step is time consuming, it is recommended to use the decomposed mesh for calculations in other directions. The boundary conditions (i.e., the direction of the imposed gradient) must be changed as desired in the *0.orig* folder.

3.1 Geometry

The microstructure of the porous media can be obtained from (1) tomography images (García-Salaberri et al. 2015a, b, 2018, 2019) or can be (2) ~~virtually generated~~ using numerical algorithms (Choi et al. 2009; Choi et al 2011; Bertei et al. 2014). Volume-averaged quantities (e.g., composition fraction) and statistical descriptors (e.g., pore size distribution, n-point correlation functions, lineal path function, and chord length function) can be used as objective variables (Pant et al. 2014). Usually, the first method provides a higher degree of fidelity to reality, even though sometimes it is difficult to determine the constituents present in the images. An example is the differentiation of carbon fibres/binder and polytetrafluoroethylene (PTFE) in GDLs due to their similar X-ray absorption properties García-Salaberri et al. (2018). ~~In contrast, the~~ virtual generation of materials allows one to overcome this issue, although the creation of realistic microstructures can be challenging in some circumstances. An example is the complex multi-component, multi-scale geometry of catalyst layers. Here, ~~fibres~~ materials with a 2D arrangement of randomly-oriented fibres similar to carbon-paper GDLs were used as an illustrative example.

279 For a specified number of cylindrical fibres, N_f , of diameter $d_f = 10 \mu\text{m}$, the
 280 steps followed for the generation of the material microstructure in a box of size
 281 $[0 - S_x, 0 - S_y, 0 - S_z]$ ($S_x = S_y = 1.5 \text{ mm}$, $S_z = 0.25 \text{ mm}$), are as follows:

- 282 1. The 3D coordinates, \mathbf{P}_1 and \mathbf{P}_2 , of the axial endpoints of the cylindrical fibres
 283 are generated randomly inside the box of material. The same z -coordinate is pre-
 284 scribed for the endpoints of each fibre ($P_{1,z} = P_{2,z}$) to achieve a 2D arrangement.
- 285 2. The x and y coordinates of the axial endpoints are translated -0.25 mm , so there is
 286 some extra material around the cropped domain of size $[0 - L_x, 0 - L_y, 0 - L_z]$
 287 that is used in the simulations ($L_x = L_y = 1 \text{ mm}$, $L_z = 0.25 \text{ mm}$). This removes
 288 edge effects.
- 289 3. Once the position of all the axial endpoints is fixed, the lateral surface of the cylin-
 290 drical fibres is triangulated using a spacing in the axial and azimuthal directions,
 291 $\Delta\mathbf{x} = (\mathbf{P}_2 - \mathbf{P}_1)/10$ and $\Delta\phi = 2\pi/20$, respectively. The unit normal vectors
 292 perpendicular to each triangle are also determined.
- 293 4. The vertices and normals of the triangles that define the lateral surface of the
 294 cylindrical fibres are written into an STL file (*facets.stl*).
- 295 5. The endcaps of the cylindrical fibres are triangulated using a spacing in the
 296 azimuthal direction equal to that used for the lateral surface ($\Delta\phi = 2\pi/20$).
- 297 6. The vertices and normals of the triangles that define the endcaps are added to the
 298 *facets.stl* file (Fig. 2).

Fibre Intersections

Fibre intersection is not explicitly taken into account in the generation of the geometry. However, when the fluid region is selected in the meshing step with *snappyHexMesh*, the mesh is adapted to the external fibres surface and the intersections among them. The solid regions inside the fibres and their intersections are removed, since they are unreachable from the fluid region.

3.2 Meshing

302 The meshing utility *snappyHexMesh* is used to mesh the pore space enclosed within
 303 the external surfaces of the domain and the complex geometry of the cylindrical
 304 fibres. A background mesh composed of cubes is first created with *blockMesh*. Then,
 305 the resulting mesh is refined with *snappyHexMesh* using the *facets.stl* file as an input.
 306 Castellated meshes were considered here, suppressing the surface snapping and layer
 307 addition steps. The resulting meshes had around 5.5 millions of cells depending on
 308 the number of fibres N_f in the sample (i.e., the porosity). The maximum number
 309 of cells achieved for intermediate porosities around $\varepsilon \approx 0.6$. An example of the
 310 generated meshes is shown in Fig. 3, including a close-up view of the refined mesh
 311 close to the fibres surface. The level of refinement used in *snappyHexMesh* was set
 312 equal to (2 4).

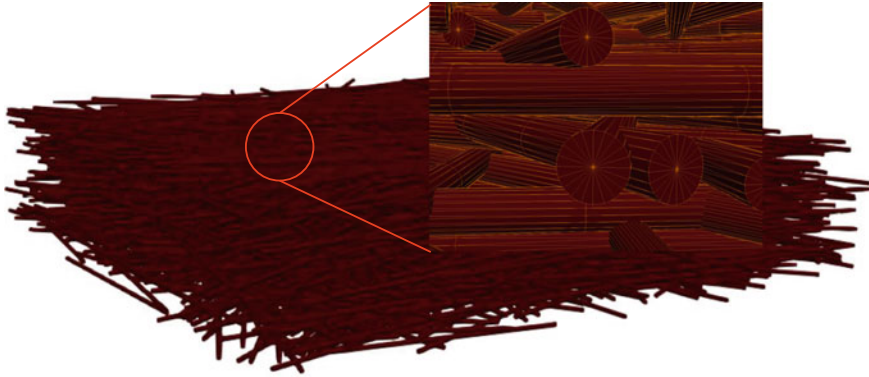


Fig. 2 Geometry composed of cylindrical fibres of $10\ \mu\text{m}$ in diameter with random orientations in the material plane (x - y plane) used to mimic the microstructure of binder-free carbon-paper GDLs. The close-up view shows the triangulated geometry

313

Practical Advice

Guidelines that should be taken into account during the meshing step are as follows:

1. Introduce a small gap slightly higher than the fibre radius near the boundary faces in the z -direction, so that the axial endpoints of the fibres do not touch the boundary faces. This facilitates the selection of a point inside the fluid region during the meshing step.
2. Set the size of the cubic cells in the background mesh similar to the fibre diameter ($d_f = 10\ \mu\text{m}$). This size ensures that the geometry of the fibres is properly captured during the refinement of the mesh with *snappyHexMesh*. No significant differences were found in the results using background cells of $5\ \mu\text{m}$ in size.

314

3.3 Solver and Postprocessing

Laplace and Navier-Stokes equations used to simulate diffusion and convection are solved in OpenFOAM with the steady-state solvers *laplacianFoam* and *simpleFoam*, respectively. A laminar *simulationType* with a Newtonian *transportModel* is set in *simpleFoam*. The remaining solver settings can be kept similar to those commonly used in OpenFOAM.

For post-processing, the *topoSet* utility is used to create a cell zone that includes the entire computational domain (i.e., the fluid region of the porous media). Then, the

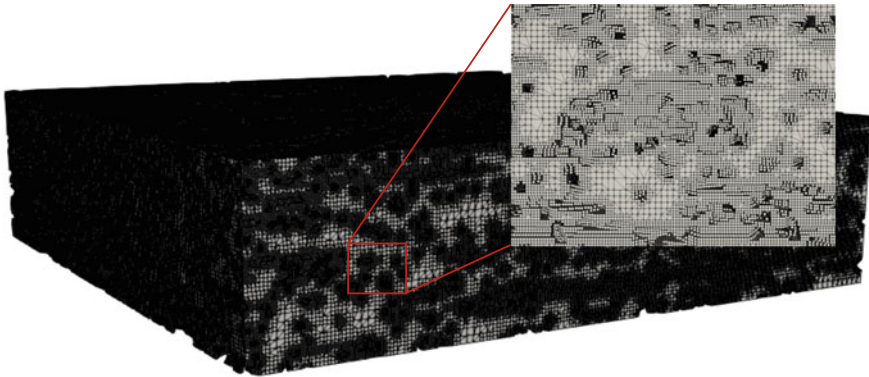


Fig. 3 Castellated mesh generated with *snappyHexMesh* using a background mesh with cubic cells of $10\ \mu\text{m}$ in size (equal to the fibre diameter, d_f). The close-up view shows the refined mesh around the fibres surface

average diffusive flux (or velocity component) in each direction is calculated using the *postProcess* utility, defining the function objects in *controlDict*. The *volAverage* operation is used for volume averaging.

Calculation of Effective Transport Properties

The pre-factors multiplying the average diffusive flux (or velocity component) in Eqs. (3) and (9) (such as the length L_i or the porosity ε) can be included as a *scale* factor in the function objects implemented in *controlDict*. The porosity can be easily determined by dividing the volume of the fluid region (provided by the *postProcess* utility) by the total volume of the domain (which is known in advance). Alternatively, the pre-factors can be introduced when the results are plotted, for example, using Python.

4 Results

In this section, the computed results for the effective diffusivity and permeability are discussed. The results in both the through- and in-plane directions are presented given the anisotropy of the generated carbon-paper GDLs. One representative direction (x -direction) is analyzed in the material plane (x - y plane). Figures 4 and 5 show some illustrative examples of the concentration distributions corresponding to effective diffusivity calculations for two different porosities (i.e., number of fibres). The pressure distributions and streamlines corresponding to permeability calculations are shown in Figs. 6 and 7. The porosities lie within the range typically observed

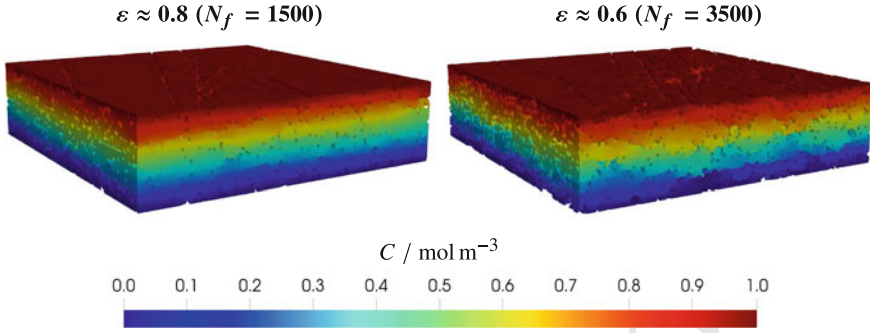


Fig. 4 Concentration fields, $C(x, y, z)$, corresponding to calculations of the through-plane effective diffusivity for two different porosities, ε (number of fibres, N_f): (left) $\varepsilon = 0.8$ ($N_f = 1500$), (right) $\varepsilon = 0.6$ ($N_f = 3500$)

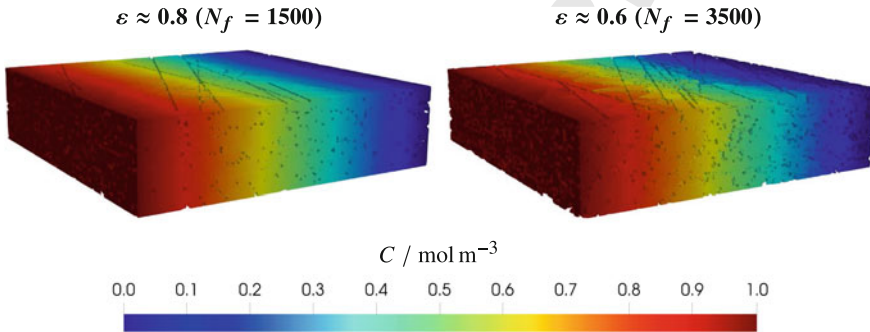


Fig. 5 Concentration fields, $C(x, y, z)$, corresponding to calculations of the in-plane effective diffusivity for two different porosities, ε (number of fibres, N_f): (left) $\varepsilon = 0.8$ ($N_f = 1500$), (right) $\varepsilon = 0.6$ ($N_f = 3500$)

338 for uncompressed ($\varepsilon = 0.8$) and mid-compressed ($\varepsilon = 0.8$) GDLs. Higher transport
 339 properties are found in the in-plane direction due to the 2D arrangement of carbon
 340 fibres, which creates larger pores in the material plane and facilitates transport in this
 341 direction.

342 A decrease of the overall diffusive flux across the material (at the same concentra-
 343 tion gradient) appears due to the decrease of the porosity and the increase of the
 344 tortuosity of transport pathways. The increment of tortuosity with porosity leads
 345 to non-linearities in the dependence between the normalized effective diffusivity,
 346 D^{eff}/D , and the porosity, ε , as established by the relationship García-Salaberri et al.
 347 (2015b):

$$348 \quad \frac{D^{\text{eff}}}{D} = \frac{\varepsilon}{\tau} \quad (10)$$

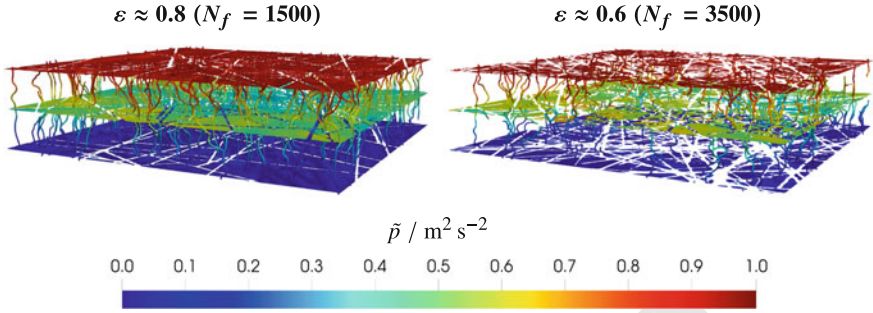


Fig. 6 Pressure fields, $\tilde{p}(x, y, z)$, and streamlines (colored by pressure level) corresponding to calculations of the through-plane absolute permeability for two different porosities, ε (number of fibres, N_f): (left) $\varepsilon = 0.8$ ($N_f = 1500$), (right) $\varepsilon = 0.6$ ($N_f = 3500$)

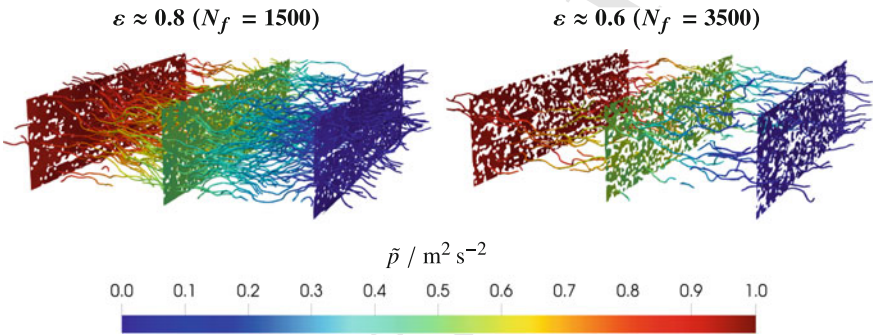


Fig. 7 Pressure fields, $\tilde{p}(x, y, z)$, and streamlines (colored by pressure level) corresponding to calculations of the in-plane absolute permeability for two different porosities, ε (number of fibres, N_f): (left) $\varepsilon = 0.8$ ($N_f = 1500$), (right) $\varepsilon = 0.6$ ($N_f = 3500$)

349 where τ is the diffusive tortuosity factor. For example, $\tau = \varepsilon^{-1/2}$ ($D^{\text{eff}}/D = \varepsilon^{1.5}$) in
 350 the traditional Bruggeman correction for porous media consisting of small, spherical
 351 solid inclusions (Bruggeman 1935; Tjaden et al. 2016).

352 Permeability of fibrous materials is influenced by hydraulic radius, porosity and
 353 tortuosity, resulting in a non-linear variation as a function of porosity Holzer et al.
 354 (2017). The Carman-Kozeny equation has previously been successfully applied to
 355 describe the variation of GDL permeability with porosity Gostick et al. (2006).

$$356 \quad K = \frac{d_f^2 \varepsilon^3}{16k_{ck}(1 - \varepsilon)^2} \quad (11)$$

357 where k_{ck} is the Carman-Kozeny constant, which is used as a fitting parameter
 358 depending on the porous medium.

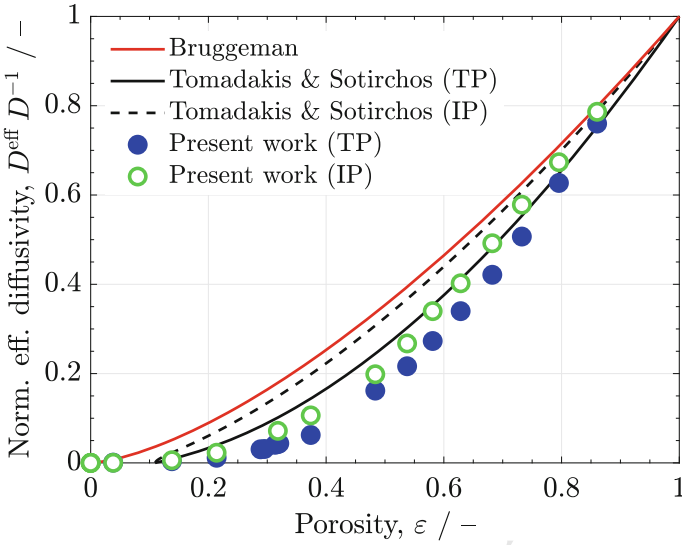


Fig. 8 Computed through- (TP) and in-plane (IP) normalized effective diffusivities, D^{eff}/D , as a function of porosity, ε . The Bruggeman correlation Bruggeman (1935), $D^{\text{eff}}/D = \varepsilon^{1.5}$, and the anisotropic random fibre model of Tomadakis and Sotirchos Tomadakis and Sotirchos (1993) (see Eq. (12)) are also shown

359 The variations of the normalized effective diffusivity and permeability as a function of porosity are shown in Figs. 8 and 9, respectively. The normalized effective
 360 diffusivity is lower than that predicted by the Bruggeman correlation, $D^{\text{eff}}/D = \varepsilon^{1.5}$,
 361 due to the more complex geometry of fibrous GDLs. Moreover, it is somewhat lower than
 362 the correlation proposed by Tomadakis and Sotirchos Tomadakis and Sotirchos
 363 (1993) for random fibre structures
 364

$$365 \quad \frac{D^{\text{eff}}}{D} = \varepsilon \left(\frac{\varepsilon - 0.11}{1 - 0.11} \right)^n \quad (12)$$

366 where $n = 0.785$ and $n = 0.521$ for the through- and in-plane directions, respectively.
 367

368 The differences between both models are ascribed to the different methodology
 369 used for the generation of the fibrous geometry. For instance, the values computed
 370 here approach those reported for binder-free GDLs, such as Freudenberg carbon
 371 paper, being $D^{\text{eff}}/D \approx 0.36$ for $\varepsilon \approx 0.65$ (Hack et al. 2020; Hwang and Weber
 372 2012).

373 The permeability is well correlated as a function of porosity using Eq. (11) with
 374 $k_{ck} = 2 - 4$. A steeper decrease of the permeability is found for porosities below
 375 $\varepsilon \lesssim 0.5$, which drops around two orders of magnitude in the range $\varepsilon = 0.1 - 0.5$
 376 compared to the ten-fold descent in the range $\varepsilon = 0.5 - 0.85$. This shows that perco-
 377 lation through the porous medium is significantly reduced at small porosities, so that
 378 important changes in permeability arise from small structural differences. Similar

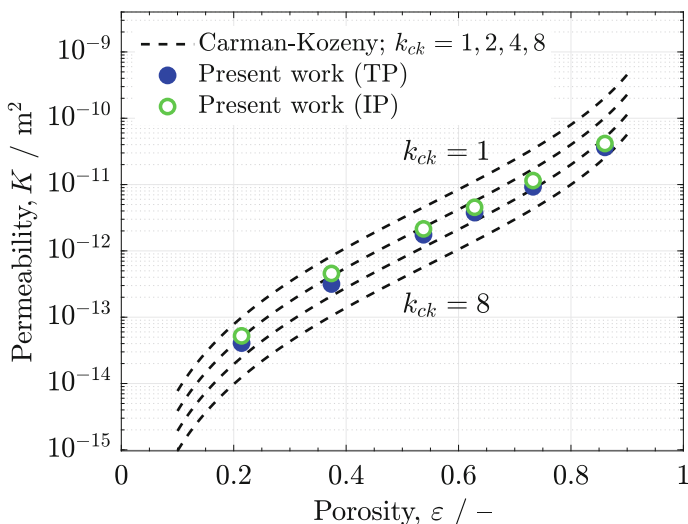


Fig. 9 Computed through- (TP) and in-plane (IP) absolute permeabilities, K , as a function of porosity, ε . The curves corresponding to various Carman-Kozeny constants, k_{ck} , are also shown (see Eq. (11))

379 results were reported in previous works examining the permeability of fibrous porous
 380 media (Tomadakis and Robertson 2005; Nabovati et al. 2009).

Effective Diffusivity and Permeability of Commercial GDLs

For a given porosity, the effective diffusivity and permeability of commercial GDLs (Toray, SGL Carbon Group and Freudenberg carbon papers) highly depends on their microstructure. In fact, the volume fraction and porosity of binder have been identified as key variables that influence the effective transport properties of GDLs (García-Salaberri et al. 2018; Mathias et al. 2003; Zenyuk et al. 2016). For example, Toray TGP-H series carbon papers show lower effective diffusivities in the through-plane direction than those computed here due to the more complex pore structure that arises from the addition of a practically non-porous binder (García-Salaberri et al. 2015a, b); $D^{\text{eff}}/D \sim 0.22 - 0.28$ (Toray TGP-H) vs. $D^{\text{eff}}/D \sim 0.45$ (present work) at $\varepsilon \approx 0.72$.

381

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