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Upscaling the elastic stiffness of foam concrete as a three-phase composite material

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Abstract

The stiffness of foam concrete depends primarily on the added porosity. Nevertheless, by performing 3D elastic numerical simulations on artificial unit cells in the frame of periodic homogenization, it is shown that describing foam concrete as a porous material is not sufficient to explain the experimental measurements of the Young modulus for added porosity higher than 40%. Indeed, introducing sand as a third phase enables to recover accurate estimates of the Young Modulus. Furthermore, for highly porous concrete foams, it is shown that the stress concentrates in thin members deprived of stiff sand particles, thus leading to a softer overall stiffness.

Keywords: foam concrete, homogenization, Young modulus, microstructure, numerical simulations

Introduction

Foam concrete, also named cellular concrete, offers many advantages as a building material. Indeed, it features a low density, a low thermal conductivity and a valuable fire and water resistance [1, 2]. Lastly, it can easily be manipulated, pumped and cast into various shapes [3]. Regarding the formulation of foam concrete, the volume fraction of void, hereby called the porosity, can be tuned to trade cost, density and insulation properties for durability, strength and stiffness [1, 5, 6, 3, 1, 7]. Typical uses include floors, trench fills, roof insulation and masonry units [8].

For lightweight concrete containing soft EPS beads formulated in [4], extending [9], it has been shown that the Young modulus does not depend on the size of the beads in the range 1mm–6.3mm, for porosities in the range 0%–50%. As a result, the Young modulus mainly depends on the porosity, exhibiting a one-to-one relationship in this domain. In addition, different mean-field homogenization schemes have been defined with the view to explain the dependence of the Young Modulus to the added porosity, thus aiming at a scientific modelling of the stiffness. For instance, the fact that the experimental measurements of the Young Modulus are compliant with the Hashin–Shtrikman upper bound [10] has been checked and the trisphere model [11, 12] has been successfully applied to provide a better estimate of the stiffness [9]. Furthermore, the differential scheme [13] ensures a good agreement on a wider range of porosity (0 ≤ p ≤ 56%) as the Young Modulus varies directly with (1 − p)² [14]. This exponent of 2 is consistent with expected and measured trends for foams where the bending is mainly attributed to the bending of the cell ribs [15, 16]. Nevertheless, for foam concretes featuring high porosities (up to 75%), power laws featuring exponents higher than 2 (E = 24GPa(1 − p)².3 [17] or E = 32.9GPa(1 − p)².8 [6]) have been fitted to experimental results, thus questioning the use of the differential scheme for porous materials.

As an alternative to Eshelby-based mean-field schemes, full-field numerical elastic simulations can be performed on an artificial unit cell of periodic composite materials to estimate the effective overall stiffness [18, 19, 20, 21]. Two steps are involved in the process. First, an artificial periodic unit cell complying with the observed microstructure of the composite material must be built. Parameters such as volume fractions, pore size distributions and minimum spacing may be accounted for in this step. Then, elastic computations help estimating the effective stiffness of the composite material by relying on the periodic homogenization theory.
The description of foam concrete as a biphasic material will be questioned by performing numerical simulations. First, concrete foam is considered as a biphasic material and the Young Modulus will be estimated for different volume fractions, pore size distributions and parameters driving the geometry of the microstructure. These estimates will be compared to experimental measurements to show that foam concrete can hardly be described as a biphasic composite material at high porosity. Then, sand is introduced as the third phase and the estimated Young Modulus is compared to the experimental measurements.

1. 3D numerical modelling

1.1. Generation of artificial unit cells

So as to perform 3D numerical simulations, unit cells complying with the morphology of the considered material must be generated. By mean of X-ray tomography, foam concrete can visibly be described as porous matrix-inclusion material, where the pores are rather spherical (Fig. 1.1). In addition, the sieve curve of the sand and the pore size distribution are such that the sand particles are smaller than the pores (Fig. 1.2). Nevertheless, there is no separation of scale between the sand and the pores. In the present section, a procedure to obtain unit cells satisfying this description is proposed.

Figure 1.1: A 6×6mm$^2$ slice extracted from a X-ray tomographic image of a foam concrete of density 0.6 and porosity 70% [17], the pixel size being 6µm. Dark areas are pores, grey uniform areas are sand particles, white dots are likely related to unhydrated cement particles.

On the one hand, for matrix-inclusion composite materials, the random sequential adsorption algorithm [22] or the Lubachevsky-Stillinger algorithm [23] are often applied to pack spherical inclusions and reach the targeted volume fractions. On the other hand, the microstructure of closed-cell foams is often modelled by using Voronoi or Laguerre diagrams. In the present work, a continuum of microstructures between these geometries is defined to model closed-cell foams of any volume fractions. First, the Lubachevsky-Stillinger algorithm [23] is adopted to produce periodic unit cells featuring thousands of spherical inclusions and volume fractions up to about 60% [24]. This algorithm is briefly recalled in the next section.

1.1.1. The Lubachevsky-Stillinger algorithm for periodic unit cells

The Lubachevsky-Stillinger algorithm [23, 25] is an event-driven algorithm designed to pack hard spherical particles in a given volume. As input, it requires the size of the unit cell and both the targeted volume fractions $\phi_i$ and size distributions for each kind of particles $i \in \text{(sand,pore)}$.

The size distribution of sand particles or pores, named the sieve curve, is defined by the volume fraction of particles $P_i(v)$ of volume lower than the volume $v$ (Fig. 1.2). For the shake
of simplicity, it is modelled as a piecewise linear function on intervals \([v_j, v_{j+1}], j \in 0, N_s\), the volume fraction of particles in these range being \(p_{i,j}\). \(N_s\) is the number of sieves. A list of targeted volumes \(v_{i,k}, k \in 0..N_u\) is randomly picked according to the sieve curve so as to fill a volume \(\sum v_{i,k} = V_{\phi_i}\), where \(V\) is the volume of the rectangular unit cell. \(N_u\) is the number of spheres of kind \(i\) to be placed in the unit cell. The Lubachevsky-Stillinger algorithm is seeded by randomly placing dots in the unit cell at time \(t = 0\). Each dot becomes a sphere as time goes by, the growth of its radius being linear. Consequently, the growth rate is set to \(a_{i,k} = (3/(4\pi v_{i,k}))^{1/3}\) so that all spheres reach their targeted volume \(v_{i,k}\) at time \(t = 1\).

The Lubachesky-Stillinger enforces that the growing spheres do not overlap. Indeed, the spheres are allowed to move at uniform velocity and collisions are handled by changing these velocities at the time of collisions. The time of the potential collision can be analytically computed since the velocities are uniform and the growth rate is linear. Consequently, the algorithm can jump from one collision to the next without time-stepping. Other potential events are encounters with the boundaries of the unit cell, where instances of the sphere must be introduced to ensure periodicity. Indeed, either the sphere hits a boundary and a new connected instance of the sphere must be created at the opposite boundary or an instance leaves the unit cell and it can be destroyed to save memory. An additional improvement consists in splitting the domain into rectangular sectors to reduce the number of potential events. Consequently, if \(N\) instances of different spheres overlap with a sector, \(N(N-1)/2 + 6N\) potential events can occur in that sector.

The output of the Lubachevsky-Stillinger algorithm specifies the center, the volume and the kind of each sphere in the unit cell. The algorithm is very efficient at packing spherical particles as it can reach volume fractions of 74% for an unimodal pore size distribution. Nevertheless, the degree of order increases monotonically with the jammed packing fraction and reaching high volume fractions (\(\geq 64.5\%\)) requires more time as the growth rates need to be reduced.

To avoid these shortcomings, an additional step allowing to reach any volume fractions of pores is introduced in the next section, though the pores become non-spherical.

1.1.2. A continuum of microstructures between spherical inclusions and Laguerre diagrams

Foams are often modelled by considering Laguerre diagrams, also called Power diagrams. These partitions of space into convex polyhedral cells are built starting from \(n\) seeds characterized by their position \(x_k\) and a scalar, named power, \(r_k\). The cell \(i\) is defined as the set of points \(z\) such that:

\[
||z - x_i||^2 - r_i^2 \leq ||z - x_k||^2 - r_k^2 \quad \forall k
\]

If all powers \(r_k\) are equal, the resulting tessellation is a Voronoi diagram. Seeding a Laguerre diagram with the centers and radii of non-overlapping spheres is a practical way to gain some control on the volume and shape of each cell of the foam (or polycrystal). Indeed, by doing so, the Laguerre cells obviously contain their corresponding spheres.
Finally, to complete the modelling of the closed cell foam, a thickness \( b_{ik} \) is associated to each solid planar member of the Laguerre diagram \([29]\), providing that the total solid volume is consistent with the targeted porosity \( \phi \).

Nevertheless, the generated microstructures features convex polyhedral pores and it clearly conflicts with the round pores of foam concrete due to the equilibrium between pore pressure and surface tensions. This issue is now to be tackled.

Starting from a sphere packing at an intermediate volume fraction \( \phi_{pore} \leq \phi \), each spherical pore can be diluted until the minimum thickness \( b_{ik} \) becomes the limiting factor. A dilation parameter \( d \) is introduced and a point \( x \) is in a pore \( i \) if and only if three conditions are met:

1. The point is in the corresponding Laguerre cell \( i \)
2. The point is not in the closed-cell foam defined by the minimum thickness \( b_{ik} \)
3. The point is in the dilated sphere:
   \[ ||x - x_i||^2 - (r_i + d)^2 \leq 0 \]  
   (2)

The value of the dilation parameter \( d \) must be set according to the targeted porosity by mean of a dichotomy algorithm. Lastly, a unique minimum thickness \( b = b_{ik} \forall i, k \) is set for all planar members so as to limit the number of parameters. Consequently, the complete procedure is driven by a couple of parameters:

- The size of the unit cell \( l \)
- The targeted volume fractions of sand \( \phi_{sand} \)
- The grain size distribution of sand \( v_{sand,j}, p_{sand,j} \)
- The targeted porosity \( \phi \)
- The pore size distribution \( v_{pore,j}, p_{pore,j} \)
- The intermediate volume fraction of pores \( \phi_{pore} \)
- The minimum thickness of planar members \( b \)

The only requirement on \( l \) is that it must be big enough for the estimated effective stiffness to be representative of the composite material. Volume fractions and size distributions can be extracted from the formulation or tomography images. Then, parameters \( \phi_{pore} \) and \( b \) are yet to be set. If the difference \( \phi - \phi_{pore} \) is significant, the microstructure deviates from well-dispersed spherical inclusions. Indeed, large planar members are created between pores which are close to one another. Then, if the minimum thickness \( b \) were null, the pores would percolate and the solid skeleton could be fragmented. Moreover, the minimum thickness \( b \) divides the matrix into two parts: the matrix is either mainly in the planar member (large \( b \)) or concentrated at the foam nodes.

It can be noticed that the modelling of highly porous foams by such a procedure is not perfect, due to the fact that the Plateau’s laws are almost always infringed: angles between the planes defining edges are likely not equal to 120° and angles between edges can be very different from \( \arccos(-1/3) \approx 109.47 \). In addition, a given non-spherical pore can feature both a flat face and curved boundaries. Since the curvature is related to the pressure drop across the interface, and since the pressure within a pore is uniform, the solid skeleton is assumed to be able to withstand some deviatoric stress during setting, through viscous effects for instance.

The resulting unit cell defines the microstructure of a periodic foam concrete (Fig. 1.3). The next section addresses the numerical computation of the effective elastic properties of this material.

### 1.2. 3D numerical simulations by the Fast Fourier Transform (FFT) algorithm

Periodic boundary conditions are applied, a mean stress is applied to the microstructure and the local strain is computed. Finally, a component of the overall stiffness is obtained by integrating the local elastic energy. By generating a unit-cell respecting periodic constrains, boundary effects are avoided and the estimated overall stiffness corresponds to that of the material defined by periodizing the unit cell. Nevertheless, the unit cell must be large enough to ensure an estimated Young modulus representative of the composite material. Different algorithms can be applied to solve the elastic problem, including the finite element method and the FFT method. The latter is preferred in the present study due to its versatility and ease of use.
Figure 1.3: Top left: spheres are packed at the intermediate volume fraction of 61% according to the considered size distribution. The corresponding Laguerre diagram is colored in blue. Top right: the distance to the nearest pore is displayed. Bottom: the resulting periodic unit cells of size 6mm×6mm×6mm at porosities of 67%, the minimum thickness being $b = 0$ (left) and $b = 24\mu m$ (right). A binning has been applied to provide an estimate of the local porosity.

1.2.1. The FFT algorithm

The FFT algorithm is recognized as one of the most powerful tool for computing the overall response of composites featuring complex periodic microstructures [32, 33, 34]. As input, the local elastic stiffness $C(x)$ must be provided at each point of a regular grid representing the unit cell. Then, either a macroscopic strain $\varepsilon$ or a macroscopic stress $\Sigma$ is applied to the unit cell. As output, the algorithm provides a value of the local strain $\varepsilon(x)$ and stress $\sigma(x)$ for each voxel $x$ of the grid as well as averages of local stresses $\bar{\sigma}$ and strains $\bar{\varepsilon}$. Finally, the elastic energy can be computed according to Hill’s lemma. Since the accelerated scheme of Eyre and Milton [33] has been pointed as the most efficient regarding the convergence rate at high contrast [34], it is applied without modification in the present study. If a macroscopic strain $\varepsilon$ is applied to the unit cell, its iteration writes:

$$
\varepsilon^{i+1} = \varepsilon^i + 2(C + C_0)^{-1} : \Gamma_0 \ast ((C_0 - C) : \varepsilon^i) - \varepsilon^i + E
$$

where $\ast$ denote a convolution and $\Gamma_0$ is the Green operator describing the strain triggered by an eigenstress in an infinite homogeneous material of an isotropic reference stiffness $C_0$. This convolution is turned into a product in the Fourier space. Indeed, the Green operator $\Gamma_0$ is explicitly known in the Fourier space for any frequency $\xi$ [32]:

$$
\hat{\Gamma}_0^{ijkl}(\xi) = \frac{1}{4\mu_0^2|\xi|^2}\left(\delta_{ij}\xi_k\xi_l + \delta_{il}\xi_j\xi_k + \delta_{jk}\xi_l\xi_i + \delta_{lj}\xi_k\xi_i\right) - \frac{\lambda_0 + \mu_0}{\mu_0(\lambda_0 + 2\mu_0)} \frac{\xi_j\xi_k\xi_l\xi_i}{|\xi|^4}
$$

where $\lambda_0$ and $\mu_0$ are the Lamé parameters of the infinite homogeneous material.
where $\lambda^0$ and $\mu^0$ are the Lamé coefficients of the isotropic reference material $\mathcal{C}_0$. The fastest convergence rate is obtained for a reference material defined as [33] [34]:

$$
k_0 = \sqrt{\frac{\max k(x)}{\min k(x)}} \quad \mu_0 = \sqrt{\frac{\max \mu(x)}{\min \mu(x)}}
$$

(5)

where $k(x)$ and $\mu(x)$ are respectively the bulk and shear modulus of the isotropic stiffness $\mathcal{C}(x)$. The complete algorithm [34] writes:

$$
\varepsilon(x) = E \quad \forall x
$$

while not converged

$$
\tau(x) = (\mathcal{C}_0 - \mathcal{C}(x)) : \varepsilon(x) \quad \forall x
$$

\hat{\tau} = \text{FFT}(\tau)

$$
\hat{\varepsilon}(\xi) = \hat{\varepsilon_0}(\xi) + \hat{\tau}(\xi) \quad \forall \xi \neq 0 \quad \hat{\varepsilon}(0) = E
$$

\hat{e} = \text{FFT}^{-1}(\hat{\varepsilon})

$$
e(x) = e(x) - \varepsilon(x) \quad \forall x
$$

\varepsilon(x)^{i+1} = \varepsilon(x) + 2(\mathcal{C}(x) + \mathcal{C}_0)^{-1} : \mathcal{C}_0 : e(x) \quad \forall x

$$
\sigma(x)^{i+1} = \mathcal{C}(x) : \varepsilon(x)^{i+1} \quad \forall x
$$

convergence tests

If a macroscopic stress $\Sigma$ is applied, the overall strain is to be updated at each iteration as [32]:

$$
E = \mathcal{C}_0^{-1} : (\Sigma - \sigma^i) + \varepsilon^i
$$

(7)

The previous algorithm requires little memory since only two fields need to be stored if the Fast Fourier Transform (FFT) is computed in place. The complexity of the iteration is that of FFT, that is $n \log(n)$ where $n$ is the size of the grid. The number of iteration needed to ensure convergence varies directly with the square root of the maximum contrast ($\max k(x) / \min k(x)$, $\max \mu(x) / \min \mu(x)$). Convergence criteria are related to the balance of the stress, compatibility of the strain and respect of the applied overall loading as defined in [34]. Among improvements of the FFT algorithms that are not considered in the present article, the use of a finite difference based Green operator [35] or a Green operator consistent with voxel-wise constant polarization fields [36, 37] may lead to more accurate local fields [35] or provide bounds on the macroscopic properties based on variational principles [36, 37].

1.2.2. Handling of pores and grey voxels

A porous material is considered: the Young modulus of the solid skeleton is set to $E_m = 24\, \text{GPa}$, its Poisson’s ratio being 0.2. Therefore, the bulk and shear modulus of the matrix respectively are $K_m = 13.33\, \text{GPa}$ and $G_m = 10.00\, \text{GPa}$. The accelerated scheme of Eyre and Milton [33] is not able to handle pores as the contrast becomes infinite. Consequently, a non-null stiffness must be associated to the pores so as to recover a finite contrast. The larger the contrast is, the softer the pore are and the more accurate the estimated overall response becomes. Nevertheless, the computation time increases as the square root of the contrast. As a consequence, a sensitivity study to the contrast is to be performed: contrasts of 100, 1000 and 10000 are tested for a given porous unit cell featuring a porosity of 63.61% (Fig. 1.4). While changing the contrast from 100 to 1000 leads to a significant decrease of the response becomes. Nevertheless, the computation time increases as the square root of the maximum contrast

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Figure 1.4: Effect of the discretization of the unit cell on the estimated stiffness. For a given cubic unit cell featuring 44 spherical inclusions and a volume fraction of 63.61%, the effective bulk modulus (left) and shear modulus (right) are estimated by using the FFT algorithm. Parameters to be tested are the discretization \(64^3, 96^3, 128^3, 144^3, 160^3\), the contrast \(100, 1000, 10000\) and the rule defining the stiffness of the grey voxels (Voigt or Reuss).

1.2.3. Representativity of the unit cell

It must be checked that the generated unit cell is a representative elementary volume (REV). Generating the unit-cell of a periodic matrix-inclusion microstructure and applying periodic boundary conditions ensures that the estimated overall response is representative of the effective stiffness of that particular microstructure. Indeed, the overall response is not plagued by boundary effects which occur as displacement or force boundary conditions are considered, as investigated in [42]. Nevertheless, if the unit cell is too small, the estimate of the stiffness can be erroneous since the unit cell does not account for the size distribution and other statistics of the material. For a maximum error of 5% on the estimated overall modulus of matrix-inclusion materials, the unit cell must be at least twice as large as the reinforcement diameter [43].

Furthermore, the statistical variability of the estimated stiffness can be plotted as a function of the size of the unit cell to justify the choice of this size [18, 19, 21, 44]. Unit cells of periodic microstructures featuring a porosity of 61% and a pore size distribution given in figure 1.2 (maximum diameter \(D = 2.21\) mm) are generated by using the Lubachevsky-Stillinger algorithm. Unit cells of sizes \(4 \times 4 \times 4\) mm\(^3\), \(6 \times 6 \times 6\) mm\(^3\) and \(8 \times 8 \times 8\) mm\(^3\) are discretized on grids of respective sizes \(80^3\), \(120^3\) and \(160^3\) so as to discard the effect due to the pixel size. Ten unit cells of each size are created so as to compute the average and the standard deviation of the estimated bulk and shear moduli. It is shown on figure 1.5 that the standard deviation is reduced as the size of the unit cell increases. Moreover, all the estimated bulk moduli and shear moduli respectively fall into ranges \((2.63\) GPa, \(2.69\) GPa) and \((1.72\) GPa, \(1.85\) GPa). Consequently, it can reasonably be assumed that the relative error on the estimated bulk and shear moduli due to representativity is less than 5% for a given unit cell of size \(6 \times 6 \times 6\) mm\(^3\) or larger.

2. Comparison between numerical estimates and experimental measurements

2.1. Foam concrete as a porous material

Various unit cells featuring different pore size distributions, different intermediate porosities and minimum thicknesses are considered.

2.1.1. Influence of the pore size distribution

For each porosity, three pore size distributions are considered:

- An unimodal pore size distribution, where all pores feature an equivalent diameter of 1.5mm.
- A bimodal pore size distribution, where half the porosity corresponds to pores of equivalent diameter 2mm and the other half to pores of diameter 0.5mm.
- A multimodal pore size distribution, given in figure 1.2 (maximum diameter \(D = 2.21\) mm).
Figure 1.5: Effect of the size of the unit cell on the estimated stiffness. For cubic unit cells of sizes $4 \times 4 \times 4$ mm$^3$, $6 \times 6 \times 6$ mm$^3$ and $8 \times 8 \times 8$ mm$^3$ and a porosity of 61%, the scaled effective bulk modulus $K/K_m$ (left) and shear modulus $G/G_m$ (right) are estimated by using the FFT algorithm. Error bars represent the standard deviation of each series of 10 unit cells.

For generating the unit cells, the intermediate volume fraction is set to 61% if the porosity is larger than 61% and the minimum thickness is set to 0. The size of the unit cells is $6 \times 6 \times 6$ mm$^3$ and these unit cells are discretized on $120 \times 120 \times 120$ grids for the FFT algorithm. For different porosities, the overall moduli estimated by the 3D numerical computations are compared to those stemming from mean field schemes in figure 2.1. It is shown that these numerical estimates comply with the Hashin-Shtrikman upper bound and could be approximated by using the differential scheme. In addition, the different pore size distributions cannot be distinguished starting from the estimated moduli, keeping in mind their limited accuracy. As a result, the effect of the pore size distribution can be described as small and non-significant. Lastly, the experimental measurements of the Young moduli of foam concrete cannot be properly approximated by any of the computed estimates. For instance, the measured Young modulus at a porosity of 67% is about 5% of that of the solid matrix, while the ratio estimated by 3D numerical simulations is about 12.7%. That ratio becomes 11.0% for the differential scheme and 19.7% for the Hashin-Shtrikman upper bound. In the next section, the intermediate porosity and the minimum width will be tuned to try to match the measured moduli.

2.1.2. Influence of the intermediate porosity and minimum thickness

The pore size distribution being set to that given in figure 1.2 (maximum diameter $D = 2.21$ mm), the effect of the intermediate porosity and minimum thickness are to be investigated. As the intermediate volume fraction becomes significantly lower than the targeted porosity, the pores tend to bundle, thus forming larger pores of complex shapes. Consequently, this clustering induces a significant decrease of the estimated stiffness (Fig. 2.2), thus getting closer to the experimental measurements of the Young modulus of foam concrete. Nevertheless, explaining the experimental measurements by introducing a single intermediate volume fraction is hardly possible. Indeed, even if the intermediate volume fraction is null, the estimated stiffness is still stiffer than that measured at high porosities. In addition the generated unit cell does not look like the tomographic image of a real sample of foam concrete. Indeed, as the intermediate porosity becomes close to zero, the resulting microstructure becomes similar to that obtained by randomly placing overlapping spheres and the pore network is almost entirely percolated. Lastly, introducing a minimum thickness raises the estimated stiffness and does not solve the mismatch between numerical estimates and measurements.

Depending on the porosity, the following conclusions arise. On the one hand, for porosities lower than 40%, the upper Hashin-Strickman bound accurately estimates the overall stiffness. In addition, there is a little dependence to the pore size distribution, to the intermediate volume fraction or the minimum thickness. Indeed, the clustering of pores only results in bigger pores without affecting the stiffness. On the other hand, for porosities higher than 40%, the upper Hashin-Strickman bound slightly overestimates the effective stiffness of the considered unit cells. Hence, the differential scheme becomes a valuable alternative. Nevertheless, the estimated stiffness proves sensitive to the intermediate porosity and the minimum thickness and the mean-field schemes should be used with care.

However, the estimated Young Modulus is significantly higher than that experimentally measured on foam concrete. Thus, introducing the sand as a third phase could improve the
Figure 2.1: The scaled effective bulk modulus $K/K_m$ (left), shear modulus $G/G_m$ (middle) and Young modulus $E/E_m$ (right) estimated by numerical simulations are compared to those stemming from the Hashin-Shtrikman upper bound and the differential scheme. Different pore size distributions are considered for the 3D numerical simulations.

Figure 2.2: The scaled effective bulk modulus $K/K_m$ (left), shear modulus $G/G_m$ (middle) and Young modulus $E/E_m$ (right) estimated by numerical simulations are compared to those stemming from the Hashin-Shtrikman upper bound and the differential scheme. Different intermediate volume fractions are considered for the 3D numerical simulations.

numerical estimates and resolve the mismatch.
2.2. Foam concrete as a three-phase composite material

2.2.1. Questioning the scale separation

Scale separation of sand and pores is not ensured. Indeed, pores and sand particles features comparable radii, as shown on tomographic images and size distributions. Even if the sand particles were much smaller than the pores, scale separation would be questionable due to the small spacing $\delta$ between the pores. The order of magnitude of the spacing between the pores can be estimated by considering the mortar as a uniform coating on the pores. The thickness $\delta/2$ of that coating must be consistent with the volume fraction of the solid skeleton, thus leading to the following equation:

$$N_{\text{pore}} - 1 \sum_{j=0} \phi_{\text{pore},j} = \Phi \sum_{j=0} N_{\text{pore}}^{-1} \phi_{\text{pore},j} \left[ 1 + \frac{4\pi \delta}{3\sqrt{2}} \right]$$

(8)

If a single diameter $d_{\text{pore}}$ is provided, this equation boils down to $\delta = d_{\text{pore}}(\phi^{-1/3} - 1)$. If 61% of pores of equivalent diameter 1mm are considered, the spacing $\delta$ is about 179$\mu$m. If 67% of pores of equivalent diameter 0.5mm are considered, this thickness decreases to about 71$\mu$m. Therefore, the characteristic spacing between the pores is comparable to the size of the smaller sand particles ($\geq 65\mu$m). Consequently, it is likely that there are little or no sand particles in the thin members of the solid skeleton, where the stress concentrates. As the cement paste is softer than the sand particles, neglecting this phenomenon is likely responsible for the overestimation of the overall stiffness encountered as foam concrete is considered as a two phase material. Therefore, sand is to be introduced as a third phase in the homogenization process.

On the contrary, for the EPS concrete formulated in [11], featuring porosities lower than 50% and EPS beads larger than 1mm, the thickness $\delta$ becomes larger than 260$\mu$m, which is comparable to the maximum diameter of the considered rounded quartz fine sand (300$\mu$m). As a consequence, the mortar can be considered as an homogeneous matrix and the differential scheme provides a valuable estimate of the Young modulus.

The present reasoning can be applied to other foam composite materials. For instance, the solid skeleton of the composite metal foam considered in [15] is made of an aluminum matrix reinforced by 22$\mu$m silicon carbide particles. The porosity is about 83% and the size of the pores is about 3mm. As a result, the spacing $\delta = 280\mu$m is much larger than the particles and the particles are clearly visible on BSE and SE micrographs of the cell walls [14]. Another example is the case of reinforced polyurethane foams dedicated to acoustic insulation [16]. The typical porosities and cell size are 95% and 350$\mu$m. The corresponding thickness is about 6$\mu$m. It is to be compared to the size of the fillers : 13$\mu$m for talc, 10$\mu$m for Zinc Borate and 33$\mu$m for Aluminum Hydroxide. The fact that Zinc Borate particles lead to a better improvement of the storage modulus than Aluminum Hydroxide particles [16] is likely related to their smaller size: more of these particles are located in the thin cell walls and ribs, where the strain concentrates during dynamic tests. Nevertheless, while talk particles are slightly bigger than the Zinc Borate particles and despite the fact that they induce smaller pores, the storage modulus is improved. As explained in [16], these particles are likely trapped in the thinner cell walls during the synthesis of the foam due to their particular hydrophobic property. Consequently, the size and microstructural locations of the particles are clearly influencing the elastic properties of the reinforced polyurethane foam, which is modelled as a three-phase material in [17].

Regarding foam concrete, the segregation of sand particles at foam nodes could be attributed to a capillary effect. Indeed, in a fresh mortar, the occurrence of a big particle in a thin wall would lead to an important curvature of the mortar-pore interface. The resulting capillary pressure would push the particle toward the nodes of the foam so as to recover smooth spherical pores or flat interfaces. This phenomenon, investigated in [18, 19, 50] (see 51 for a review), led to the definition of a confinement parameter $\lambda$ which is close to the ratio $d_p/\delta$, where $d_p$ is the diameter of the solid particles. The confinement parameter for a porosity of 61%, a pore diameter of 1mm and a particle diameter of 0.5mm is about $\lambda \approx 1.8$. Therefore, the sand particles are likely trapped at the nodes of the considered foam concrete.

The segregation of sand particles at foam nodes is made obvious by analysing the 3D tomographic images (Fig. 1.1 1). Nevertheless, distinguishing the siliceous sand particles from the cement paste by processing the image is hardly achievable because the X-ray linear attenuations of both materials are similar within the considered X-ray energy range [24, 63]. As an alternative, the image is treated so as to segment the volume of the solid skeleton that could be occupied by spherical sand particles larger than a given volume (Fig. 2.3). To this
end, a median filter is first applied to reduce noise (Fig. 2.3a) followed by an Ostu threshold to segment the solid skeleton (Fig. 2.3b). The segmented volume fraction for the foam concrete of porosity 70% is 42%. Therefore, the segmented solid skeleton likely overestimates the extend of the actual solid skeleton. Then an opening morphological filter is applied so as to define the volume that could be occupied by sand grains larger than a given diameter. Hence, the structuring element of the morphological filter is a ball featuring the considered diameter. As the diameter of the ball increases from 60µm to 144µm, the remaining volume accessible to sand grains significantly decreases from 37% to 24%. Hence, a significant volume of the solid skeleton is made of thin members of cement paste deprived of sand grains, since the considered sieve curve starts at an equivalent diameter of 250µm (Fig. 2.3d). All these treatments are performed on the 3D image using the Insight Segmentation and Registration Toolkit (ITK) [54, 55]. Unsurprisingly, right after thresholding, the solid skeleton is found to be made of a single continuous component, apart from some small spurious components near the edges of the image. On the contrary, once the thin members deprived of sand grains are removed, the solid skeleton becomes disconnected. Indeed, at diameter 144µm, in an abstract of size 1.8mm×3mm×3mm (Fig. 2.3), the three largest components respectively occupy volume fractions of 7.8%, 3.5% and 3.4% and components disconnected from the edges of the domain reach a volume fraction of 1.1%. It proves that these thin members are essential to the stiffness of the foam.

Figure 2.3: 3mm×3mm slices of a treated tomographic abstract of size 1.8mm×3mm×3mm. First a median filter is applied, then an Ostu threshold filter, an opening by a ball and a labelling of the components. As the diameter of the ball increases, the colored components depicting the volume potentially occupied by sand grains of diameter bigger than that of the ball become disconnected.

2.2.2. 3D numerical estimates of the stiffness

Both sand particles and pores are now accounted for. The volume fraction of sand in the mortar is $f_{s/m} = 35\%$ and the size distribution of sand is plotted in figure 2.2 (maximum diameter $D = 2\text{mm}$), along with that of pores (maximum diameter $D = 2.21\text{mm}$). Hence,
\( \phi \) being the final porosity, the volume fraction of sand in foam concrete is \( f_{s/m}(1 - \phi) \). The total volume fraction of particles and pores is therefore \( f_{p/m}(1 - \phi) + \phi \). An intermediate volume fraction is to be introduced for the pores: it is set to 61% = \( f_{p/m}(1 - \phi) \) and sand particles are not modified during the dilation process, thus preserving their spherical shape.

Now that the sand and the cement paste are distinguished, it is required to set a stiffness for each of these phases. The Young modulus and Poisson’s ratio of the quartz sand particles are respectively set to \( E_s = 95.5 \text{GPa} \) and \( \nu_s = 0.078 \) [56] (see also [57]). The Poisson’s ratio of the cement paste is set to 0.2 and its Young modulus to 14.1GPa so as to recover a Young modulus of \( E_m = 24 \text{GPa} \) for the mortar by using a Mori-Tanaka scheme. The bulk modulus and shear modulus estimated by the Mori-Tanaka scheme respectively are \( K_m = 12.5 \text{GPa} \) and \( G_m = 10.2 \text{GPa} \). Such a Young modulus of 14.1GPa for a cement paste of water-to-cement ratio of 0.38 is very low compared to that measured on pure cement paste of similar age and water-to-cement ratios, around 20–25GPa [58]. The difference can partly be attributed to the mismatch between static moduli and dynamic moduli, since dynamic moduli are measured in [58] while the moduli considered in the present study are static moduli measured by performing three loading cycles between 0% and 30% of the compressive strength. Furthermore, it may be attributed to a slow hydration and curing conditions: the cement (CEM I 42.5N) is rather coarse as it features a Blaine fineness of 3200 cm\(^2\)/g and the samples are unmolded at 6-8 hours and stored in open air. However, the occurrence of an ITZ or microcracking of the cement paste in the mortar are not visible on tomographic images of foam concrete. Lastly, though the Young modulus of the mortar estimated by using the Mori-Tanaka scheme is set so as to be consistent with its experimental measurement, the Poisson’s ratio is estimated at 0.18 without experimental calibration.

The stiffness of grey voxels containing three phases must be set according to a mixing rule for elastic linear simulations using the FFT algorithm. First, the cement paste and sand grains are merged using a Reuss bound, as performed in [38]. Then, this mix is merged with the pores using a Voigt bound, as justified in section 1.2.4.

First, the minimum thickness is set to zero. The porosity being set, the overall estimated moduli are softer that those estimated by assuming the mortar to be homogeneous (Fig. 2.1). As a result, the estimates of the Young modulus accounting for sand grains are more consistent with the experimental measurements of the Young moduli on foam concrete samples. In addition, the artificial unit cells seem closer to the tomographic images and the thin members squeezed between pores are actually deprived of sand grains (Fig. 1.1,2.5). Displaying components of the local stress unveils stress concentrations patterns (Fig. 2.5). On the one hand, at low porosity (29%), the stress concentrates in the sand particles, thus leading to an efficient stiffening of the concrete foam. On the other hand, at high porosity (67%), the stress concentrations are driven by the geometry of the solid skeleton. Indeed, the stress concentrates in thin members made of soft cement paste, thus explaining the lower overall estimated moduli: these thin members acting as soft microscopic pillars. The occurrence of large stress concentrations at high porosities could also explain the reduced compressive strength of concrete foam [17].

Finally, the estimated moduli are found to be sensitive to the minimum thickness of the members for high porosities. Indeed, introducing a minimum thickness of 24\( \mu \text{m} \) induces a 16% increase of the estimated Young moduli at 67% porosity. Nevertheless, the tomographic images do not feature such a measurable minimum thickness. In addition, keeping a null minimum thickness leads to the most accurate estimates of the overall moduli Young moduli.

2.2.3. Discussion

The proposed model of foam concrete does not account for the potential occurrence of an interfacial transition zone (ITZ) around the sand particles that might affect the elastic stiffness [59, 60]. Indeed, the typical width of the ITZ is similar to the size of the cement grains (50\( \mu \text{m} \)), that is comparable to the voxel size (6mm/120) in the elastic simulations: representing the ITZ in elastic simulations at the scale of mortars would likely result in reducing the stiffness of grey voxels around the sand particles [61], replacing each sand particle by a bigger softer particle incorporating the ITZ [62] or handle the ITZ as an imperfect interfaces [63] since the discretization is too coarse to represent the ITZ as an effective layer [63, 58]. The occurrence of an ITZ is often attributed to the wall effect: based on geometrical considerations, the volume fraction of big cement particles near sand grains must be very low. Hence, the water to cement ratio is these areas is supposed to be higher, thus triggering a lower water to cement ratio far from the sand grains. In the case of foam concrete featuring a high porosity, if the ITZ were limited to sand particles, the thin members of the foam
Figure 2.4: The scaled effective bulk modulus $K/K_m$ (left), shear modulus $G/G_m$ (middle) and Young modulus $E/E_m$ (right) estimated by numerical simulations are compared to those stemming from the Hashin-Shtrikman upper bound and the differential scheme for a porous material. The cement paste, the sand particles and the pores are distinguished for the 3D numerical simulations. The moduli estimated by the Mori-Tanaka scheme, using the cement paste as the matrix, are also displayed.

Figure 2.5: Top: Generated three-phase artificial microstructures are discretized on regular grids for porosities of 29% (left) and 67% (right). Pores are black, sand grains are white and the cement paste is grey. Bottom: the component $\sigma_{xx}(x)$ of the local stress is displayed for a hydrostatic load $\Sigma_{xx} = \Sigma_{yy} = \Sigma_{zz} = 1$.

deprived of sand particles would become stiffer. Since the stress concentrates in these thin members, the overall stiffness of the foam concrete could be improved.

On the contrary, the ITZ may not be limited to sand particles. Indeed, if it is admitted that the capillary forces are strong enough to exclude small sand particles from the thin members of the foam, this phenomenon might also affect big cement particles. It would result to a capillary-induced ITZ around the pores: the water to cement ratio in the thin member of the foam would be higher, thus lowering the stiffness where the stress concentrates. As a consequence, the overall stiffness would be lower.
3. Conclusion

The present study is dedicated to foam concrete, showing that the scales of sand, pores and cement paste can hardly be uncoupled to upscale the elastic stiffness. While 3D numerical simulations and mean field schemes provide consistent estimates of the overall stiffness of a biphasic composite material made of spherical pores embedded in a matrix, the comparison to some moduli measured on foam concrete samples is puzzling, notably for high porosities. To resolve the mismatch, foam concrete can be modelled as a three-phase material by considering the cement paste, the sand and the air voids. Indeed, at high porosity, the width of the mortar members becomes so small that the big sand particles are expected to be segregated at the foam nodes. Since the stiffness of the foam is mainly driven by the ribs and walls where the strain concentrates, the effective stiffness is lower than the one expected if the mortar was homogeneous. The thickness of the foam members can be estimated according to the equation \[S_{\text{sand}} \text{and compared to the maximum diameter of the sand to decide whether considering the mortar as homogeneous is acceptable. This procedure could be extended to various porous materials. Regarding the formulation of foam concrete, the present study unveils that a fine sand is often required to ensure an homogeneous mortar and an optimized stiffness.

The elastic stiffness is not the only physical property of interest as concrete foam is considered. Whether the sand is to be considered or not as other physical linear phenomena (electrical or thermal conductivity) are investigated remain unanswered. Indeed, the pores and the sand may both provide insulation and contrasts of electric or thermal conductivities can be different from the contrast of elastic stiffnesses. The numerical simulations can be adapted to provide estimates of the overall electrical of thermal conductivities if corresponding conductivities are provided for the air pores, the cement paste and the sand particles. Finally, the increase of local stress concentrations as porosity increases could be related to a lower overall strength of foam concrete.
