



Adaptive numerical homogenization of nonlinear diffusion problems

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1 **ADAPTIVE NUMERICAL HOMOGENIZATION OF NON-LINEAR**
2 **DIFFUSION PROBLEMS***

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5 **Abstract.** We propose an efficient numerical strategy for simulating fluid flow through porous
6 media with highly oscillatory characteristics. Specifically, we consider non-linear diffusion models.
7 This scheme is based on the classical homogenization theory and uses a locally mass-conservative
8 formulation. In addition, we discuss some properties of the standard non-linear solvers and use an
9 error estimator to perform a local mesh refinement. The main idea is to compute the effective pa-
10 rameters in such a way that the computational complexity is reduced without affecting the accuracy.
11 We perform some numerical examples to illustrate the behaviour of the adaptive scheme and of the
12 non-linear solvers. Finally, we discuss the advantages of the implementation of the numerical ho-
13 mogenization in a periodic media and the applicability of the same scheme in non-periodic test cases
14 such as SPE10th project.

15 **Key words.** Flow in porous media, homogenization, upscaling, adaptive computations, non-
16 linear solvers, MFEM.

17 **AMS subject classifications.** 35K55, 76S05, 76M50, 76M10.

18 **1. Introduction.** Non-linear parabolic problems are encountered as mathemat-
19 ical models for several real life applications. Examples in this sense are partially
20 saturated flow in porous media, non-steady filtration, and reaction-diffusion systems.
21 Realistic applications often involve heterogeneous media, which translate into highly
22 oscillatory coefficients and non-linearities. Letting Ω^ε be a bounded, possibly perfor-
23 ated domain in \mathbb{R}^d ($d = 2, 3$) with Lipschitz boundary and $T > 0$ a maximal time
24 with $\Omega_T^\varepsilon := \Omega^\varepsilon \times (0, T]$, we consider the non-linear diffusion equation

25 (1.1) $\partial_t b^\varepsilon(\mathbf{x}, p^\varepsilon(\mathbf{x}, t)) - \operatorname{div}(\mathbf{K}^\varepsilon(\mathbf{x}) \nabla p^\varepsilon(\mathbf{x}, t)) = f^\varepsilon(\mathbf{x}, t), \quad \text{in } \Omega_T^\varepsilon,$

26 with suitable initial and boundary conditions.

27 In this setting, ε is a positive small parameter and denotes the scale separation
28 between the micro-scale (e.g the scale of pores in a porous medium) and the macro-
29 scale (e.g the Darcy scale, the scale of simulation in case of heterogeneous media).
30 With the superscript $0 < \varepsilon \ll 1$ we indicate that the quantities involve highly oscil-
31 latory features and the medium is considered highly heterogeneous. Equation (1.1)
32 can for example represent the non-dimensional Richards' equation after applying the
33 Kirchhoff transformation and without taking into account gravity effects (see [7]). In
34 this case, the primary unknown $p^\varepsilon(\mathbf{x}, t)$ is the transformation of the fluid pressure. For
35 simplicity $p^\varepsilon(\mathbf{x}, t)$ will be called *pressure* in what follows. The given data include the
36 source f^ε , the absolute permeability matrix \mathbf{K}^ε and the volumetric fluid saturation
37 b^ε , which is a given function of p^ε .

38 The development of numerical methods capturing the interaction between scales
39 relies on high computational costs. The use of classical schemes over fine-scale meshes

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40 has often unreachable requirements. To capture the oscillations in the medium the
 41 required mesh size would be very small compare to ε . In this sense, the standard
 42 numerical methods will either fail or become inefficient.

43 In consequence, there is a significant set of techniques for handling simulations
 44 that involve two or more scales in space and time. During the last years, approaches
 45 like the multi-scale finite-volume (MSFV), the algebraic dynamic multilevel (ADM)
 46 and the heterogeneous multi-scale (HMM) methods are becoming more and more
 47 relevant. Concretely, the MSFV and ADM methods proposed in [21, 22, 25] aim to
 48 solve problems involving different scales by incorporating the fine-scale variation into
 49 the coarse-scale operators. The multi-scale finite volume method (FVM) proposed
 50 in [22] is extended in [21] by including a dynamic local grid refinement method to
 51 provide accurate and efficient simulations employing fine grids only where needed.
 52 We highlight that the MSFV and ADM use a section of the fine-scale feature to
 53 construct the macro-scale solution without estimations of the macro-scale parameters.
 54 On the other hand, the HMM (see [3, 40]) relies on coupled macro and micro-scale
 55 solvers using homogenization (see [26]). This method takes advantage of the scale
 56 separation and is based on the numerical approximation of the macro-scale data.
 57 In [1, 2, 3] ideas on how to manage different scales in an efficient computational
 58 way are developed, using the standard finite element method (FEM). Further, the
 59 numerical computations using finite difference and discontinuous Galerkin method
 60 also demonstrate the potential of this framework in [18, 40].

61 Recently, many other works have proposed improved multi-scale methods to sim-
 62 ulate non-linear single-phase and multi-phase flow. Among the recent literature, we
 63 emphasize the approaches in [5, 6, 23, 37, 41]. An Enhanced Velocity Mixed element
 64 method is proposed in [41] to deal with non-matching, multi-block grids and cou-
 65 ple micro and macro scale domains. In the same line of research [6, 23, 37] give a
 66 computational strategy for the multi-scale dynamics over non-matching grids using
 67 mesh refinement and enriched multi-scale basis functions. In [5], the homogenization
 68 theory is combined with domain decomposition to obtain locally effective parameters
 69 and solve macro-scale problems.

70 In this paper, we develop a locally mass-conservative scheme that computes the
 71 homogenized permeability field of (1.1) over coarse meshes. In contrast with the pa-
 72 pers mentioned before, we avoid the general grids and only use a-posteriori estimators
 73 on the macro-scale solvers. We propose a combination of techniques supported in the
 74 theoretical framework of the homogenization (see [26]) for non-linear parabolic equa-
 75 tions. This strategy relies on the solution of micro-scale cell problems to calculate
 76 averaged parameters that one uses in a macroscopic solver. The focus of this work
 77 is to construct an efficient numerical strategy to approximate the solution of a non-
 78 linear and homogenized macro-scale model. It is important to remark that, despite
 79 the assumptions of periodicity that are needed in the classical homogenization the-
 80 ory, the advantages of this upscaling technique can be exploited even in the case of a
 81 non-periodic medium.

82 We apply the backward Euler (BE) method for the time discretization and the
 83 mixed finite element method (MFEM) for the spatial discretization. In order to solve
 84 the fully discrete formulation of (1.1), non-linear solvers are required. We discuss
 85 the applicability of classical iterative solvers like Newton or Picard (see [8, 17]) and we
 86 detail the formulation of a robust fixed point method called L-scheme proposed in
 87 [31].

88 This linearization procedure has the advantage of being unconditionally conver-
 89 gent. More exact, the convergence of the L-scheme is neither affected by the initial

90 guess nor by the mesh size. Nevertheless, the convergence rate of the L-scheme is
 91 only linear and therefore slower compared to the Newton scheme (see [33]). We men-
 92 tion [28] for an approach combining the L and the Newton schemes in an optimized
 93 way. There, the L-scheme is applied to provide a suitable initial point for the Newton
 94 scheme. We use this strategy to improve the convergence of the scheme up to the
 95 quadratic convergence. We also refer to [29] for a modified L-scheme featuring im-
 96 proved convergence (compared to the L-scheme) and scalability properties (compared
 97 to Newton and Picard).

98 For time dependent problems the idea of adaptive meshes is very useful to lo-
 99 calize the computational error. On the other hand, reaching finer meshes becomes
 100 computationally expensive because it requires extra calculations of the macro-scale
 101 parameters. The finer the mesh for the upscaled model, the higher the computational
 102 effort as the effective parameters need to be computed in more points, thus more cell
 103 problems need to be solved. For this reason, we present an a-posteriori estimator
 104 that indicates when the numerical solution and the effective parameters should be re-
 105 computed. With this strategy we aim to control the convergence rate of the numerical
 106 scheme and to avoid unnecessary computations of the local problems.

107 The main idea in this work is to exploit the advantages of the homogenization
 108 theory, adaptive mesh refinement and linearization procedures to obtain an efficient
 109 multi-scale solver for non-linear parabolic problems. The paper is organized as follows.
 110 In [Section 2](#) the details of the model, the geometry and the discrete formulation are
 111 given and the necessary assumptions are stated. [Section 3](#) gives a brief summary of
 112 the standard procedure of the homogenization for a parabolic case in a periodic porous
 113 media. There we also state the mixed and fully discrete formulation of the upscaled
 114 problem. In [Section 4](#) the adaptive technique based on a-posteriori error estimators
 115 is introduced and in [Section 5](#) the L-scheme is described. Finally, in [Sections 6](#)
 116 and [7](#) we discuss numerical tests in the quasi-periodic and non-periodic case and
 117 some conclusions.

118 **2. Model formulation and discretization.** We consider the following non-
 119 linear parabolic problem, which appears on models of single-phase flow through a
 120 porous media

$$\begin{aligned}
 121 \quad (2.1) \quad & \partial_t b^\varepsilon(\mathbf{x}, p^\varepsilon(\mathbf{x}, t)) - \operatorname{div}(\mathbf{K}^\varepsilon(\mathbf{x}) \nabla p^\varepsilon(\mathbf{x}, t)) = f^\varepsilon(\mathbf{x}, t), & \text{in } \Omega_T^\varepsilon, \\
 & p^\varepsilon(\mathbf{x}, t) = 0, & \text{on } \partial\Omega_T^\varepsilon, \\
 & p^\varepsilon(\mathbf{x}, 0) = p_I(\mathbf{x}), & \text{in } \Omega^\varepsilon.
 \end{aligned}$$

122 Here Ω^ε is a bounded domain in \mathbb{R}^d ($d = 2, 3$) with boundary $\partial\Omega^\varepsilon$. We denote
 123 $\Omega_T^\varepsilon := \Omega^\varepsilon \times (0, T]$ and let \mathbf{n} being the unit normal pointing outwards of Ω^ε . Using
 124 the superscript $\varepsilon > 0$ we emphasize on the fact that rapidly oscillating characteristics
 125 are involved. For example, the domain either involves characteristics changing within
 126 ε -sized regions, or it may include perforation (like a porous medium).

127 Throughout this paper we use common notations from the functional analysis.
 128 By $L^p(\Omega^\varepsilon)$ we mean the space of the p -integrable functions with the usual norm.
 129 We let $\langle \cdot, \cdot \rangle$ represent the inner product on $L^2(\Omega^\varepsilon)$. For defining a solution in a weak
 130 sense we use the space $H_0^1(\Omega^\varepsilon) = \{p \in H^1(\Omega^\varepsilon) \mid p = 0 \text{ on } \partial\Omega^\varepsilon\}$ with $H_0^{-1}(\Omega^\varepsilon)$ being
 131 its dual.

132 We make the following assumptions:

133 (A1) The function $b^\varepsilon(\mathbf{x}, \cdot)$ is non-decreasing, and $b^\varepsilon(\cdot, 0) = 0$.

134 (A2) The function $b^\varepsilon(\mathbf{x}, \cdot)$ is Hölder continuous: Constants $\alpha \in (0, 1]$ and $L_b > 0$

exist such that

$$|b^\varepsilon(\mathbf{x}, p_1) - b(\mathbf{x}, p_2)| \leq L_b |p_1 - p_2|^\alpha,$$

for all $\mathbf{x} \in \Omega^\varepsilon$ and $p_1, p_2 \in \mathbb{R}$.

(A3) The permeability function $\mathbf{K}^\varepsilon : \Omega^\varepsilon \rightarrow \mathbb{R}^{d \times d}$ is symmetric and continuous for all $\mathbf{x} \in \Omega^\varepsilon$. Further, the constants $\beta, \lambda > 0$ exist such that

$$\beta \|\psi\|^2 \leq \psi^\top \mathbf{K}^\varepsilon(\mathbf{x}) \psi \leq \lambda \|\psi\|^2 \quad \text{for all } \psi \in \mathbb{R}^d \text{ and } \mathbf{x} \in \Omega^\varepsilon.$$

(A4) The initial data satisfies $p_I \in L^\infty(\Omega_T^\varepsilon)$ and the source term is $f^\varepsilon \in L^\infty(\Omega_T^\varepsilon)$.

A weak solution for the problem (2.1) is defined as

DEFINITION 2.1. A function p^ε is called a weak solution of (2.1) if $\partial_t b^\varepsilon(\cdot, p^\varepsilon) \in L^2(0, T; H_0^{-1}(\Omega^\varepsilon))$, $p^\varepsilon \in L^2(0, T; H_0^1(\Omega^\varepsilon))$ and for all $\xi \in L^2(0, T; H_0^1(\Omega^\varepsilon))$ it holds

$$\int_0^T \langle \partial_t b^\varepsilon, \xi \rangle_{H_0^{-1}(\Omega^\varepsilon) \times H^1(\Omega^\varepsilon)} dt + \int_0^T \langle \mathbf{K}^\varepsilon(\mathbf{x}) \nabla p^\varepsilon, \nabla \xi \rangle dt = \int_0^T \langle f^\varepsilon, \xi \rangle dt.$$

We refer to [4] for the existence and uniqueness of the weak solution of the above problem. As a consequence one can also prove that $b^\varepsilon(\mathbf{x}, \cdot) \in L^\infty(0, T; L^1(\Omega^\varepsilon))$ (see [4]).

2.1. Mixed formulation. In order to construct a robust and locally conservative scheme we consider the mixed formulation of (2.1). By defining $\mathbf{u}^\varepsilon(\mathbf{x})$ as the Darcy velocity, the unknowns $(p^\varepsilon, \mathbf{u}^\varepsilon) \in L^2(0, T; H_0^1(\Omega^\varepsilon)) \times L^2(0, T; H(\text{div}, \Omega^\varepsilon))$ satisfy

$$(2.2) \quad \begin{aligned} \partial_t b^\varepsilon(\mathbf{x}, p^\varepsilon(\mathbf{x}, t)) + \text{div}(\mathbf{u}^\varepsilon(\mathbf{x}, t)) &= f^\varepsilon(\mathbf{x}, t), & \text{in } \Omega_T^\varepsilon, \\ \mathbf{u}^\varepsilon(\mathbf{x}, t) &= -\mathbf{K}^\varepsilon(\mathbf{x}) \nabla p^\varepsilon(\mathbf{x}, t), & \text{in } \Omega_T^\varepsilon, \\ p^\varepsilon(\mathbf{x}, t) &= 0, & \text{on } \partial\Omega_T^\varepsilon, \\ p^\varepsilon(\mathbf{x}, 0) &= p_I, & \text{in } \Omega^\varepsilon, \end{aligned}$$

with $H(\text{div}, \Omega^\varepsilon) = \{\mathbf{v} \in [L^2(\Omega^\varepsilon)]^d \mid \text{div}(\mathbf{v}) \in L^2(\Omega^\varepsilon)\}$. It can be proved that the mixed variational formulation (2.2) is equivalent to the conformal formulation (2.1). We refer to [32] for the proof in both continuous and semi-discrete cases.

2.2. The non-linear fully discrete problem. To define the discrete problem we let $\mathfrak{T}_{h^\varepsilon}$ be a triangular partition of the domain Ω^ε with elements \mathcal{T} of diameter $h_\mathcal{T}^\varepsilon$ and $h^\varepsilon := \max_{\mathcal{T} \in \mathfrak{T}_{h^\varepsilon}} h_\mathcal{T}^\varepsilon$ such that $h^\varepsilon \ll \varepsilon$. Further, $0 = t_0 \leq t_1 \leq \dots \leq t_N = T$, $N \in \mathbb{N}$ is a partition of the time interval $[0, T]$ with constant step size $\Delta t = t_{i+1} - t_i$, $i \geq 0$. For the discretization of the flux \mathbf{u}^ε we consider the lowest-order Raviart-Thomas space $V_{h^\varepsilon} := \mathcal{RT}_0(\mathfrak{T}_{h^\varepsilon})$ and for the pressure p^ε we use the discrete subspace of piecewise constant functions W_{h^ε} (see [10]):

$$\begin{aligned} W_{h^\varepsilon} &:= \{q \in L^2(\mathfrak{T}_{h^\varepsilon}) \mid q \text{ is constant on each element } \mathcal{T} \in \mathfrak{T}_{h^\varepsilon}\}, \\ V_{h^\varepsilon} &:= \{\mathbf{v} \in H_0(\text{div}, \mathfrak{T}_{h^\varepsilon}) \mid \mathbf{v}|_{\mathcal{T}} = \mathbf{a} + b\mathbf{x} \text{ for all } \mathcal{T} \in \mathfrak{T}_{h^\varepsilon}, \mathbf{a} \in \mathbb{R}^d, b \in \mathbb{R}\}. \end{aligned}$$

The fully discrete mixed finite element formulation of the problem (2.1) is

Problem (PM_n^ε). For a given $((p^\varepsilon)_{h^\varepsilon}^{n-1}, (\mathbf{u}^\varepsilon)_{h^\varepsilon}^{n-1}) \in W_{h^\varepsilon} \times V_{h^\varepsilon}$ and $n \geq 1$. Find $(p^\varepsilon)_{h^\varepsilon}^n \in W_{h^\varepsilon}$ and $(\mathbf{u}^\varepsilon)_{h^\varepsilon}^n \in V_{h^\varepsilon}$ such that for any $q \in W_{h^\varepsilon}$ and $\mathbf{v} \in V_{h^\varepsilon}$ there holds

$$(2.3) \quad \begin{aligned} \langle b^\varepsilon(\cdot, (p^\varepsilon)_{h^\varepsilon}^n) - b^\varepsilon(\cdot, (p^\varepsilon)_{h^\varepsilon}^{n-1}), q \rangle + \Delta t \langle \text{div}((\mathbf{u}^\varepsilon)_{h^\varepsilon}^n), q \rangle &= \Delta t \langle f^\varepsilon, q \rangle, \\ \langle [\mathbf{K}^\varepsilon]^{-1}(\mathbf{u}^\varepsilon)_{h^\varepsilon}^n, \mathbf{v} \rangle - \langle (p^\varepsilon)_{h^\varepsilon}^n, \text{div}(\mathbf{v}) \rangle &= 0. \end{aligned}$$

169 Where $(p^\varepsilon)_{h^\varepsilon}^0$ is the L^2 -projection of the initial condition p_I over the mesh $\mathfrak{T}_{h^\varepsilon}$. For
 170 simplicity we omit writing the \mathbf{x} argument in $b^\varepsilon(\mathbf{x}, p^\varepsilon)$, which becomes now $b^\varepsilon(p^\varepsilon)$.
 171 Here we assume that a solution to the discrete problem (2.3) exists and is unique. For
 172 details about this we refer the reader to [32, 35].

173 Note that the discrete problem (2.3) is non-linear, therefore a non-linear solver is
 174 needed. This is detailed in Section 5.

175 **3. Two-scale approach: Periodic case.** Consider Ω^ε as a bounded domain
 176 in \mathbb{R}^d ($d = 2, 3$) with Lipschitz boundary. We call *micro-scale* the region Y where
 177 the parameters change rapidly. In other words, the parameters and non-linearities
 178 take different values inside of Y (see Figure 1). In the extreme case, the micro-scale
 179 Y can be viewed as a perforated region with a pore space and a solid grain (see e.g
 180 [26]). Here we give the ideas for a non-perforated domains but this can be adapted
 181 straightforwardly to perforated ones.

182 At the micro-scale Y and the macro-scale Ω^ε we assume characteristic lengths ℓ
 183 and L respectively. The factor $\varepsilon := \frac{\ell}{L}$ denotes the scale separation between the two
 184 scales. To identify the variations at the micro-scale we define a fast variable $\mathbf{y} := \frac{\mathbf{x}}{\varepsilon}$.
 185 For each macro-scale point $\mathbf{x} \in \Omega^\varepsilon$ we use one micro-scale cell Y to capture the fast
 186 changes in the parameters.

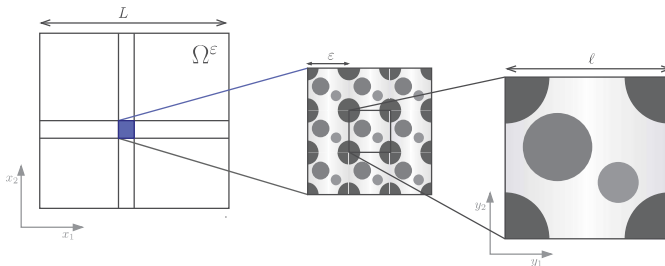


FIG. 1. Two-scales scheme. Zoom in to the pore structure in \mathbb{R}^2 where typical length sizes and the multi-scale variables are indicated. (See [11]).

187 In the non-dimensional setting, we assume that Ω^ε can be written as the fi-
 188 nite union of the local cells $Y = [0, 1]^d$. To be specific, we let $\vec{i} \in \mathbb{Z}^d$ and $\Omega^\varepsilon =$
 189 $\cup \left\{ \varepsilon(\vec{i} + Y) \mid \vec{i} \in \mathcal{I}_\varepsilon \right\}$ for some set of vector indices \mathcal{I}_ε and the outer boundary of Ω^ε
 190 is $\partial\Omega^\varepsilon$.

191 In order to formulate a homogenized problem, we make the following assumptions:

192 (B1) There exists a function $b : \Omega^\varepsilon \times \mathbb{R}^d \times \mathbb{R} \rightarrow \mathbb{R}$ such that $b^\varepsilon(\mathbf{x}, p^\varepsilon) := b(\mathbf{x}, \frac{\mathbf{x}}{\varepsilon}, p^\varepsilon)$
 193 and $b(\mathbf{x}, \cdot, p^\varepsilon)$ is Y -periodic.

194 (B2) There exists a function $\mathbf{K} : \Omega^\varepsilon \times \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$ such that $\mathbf{K}^\varepsilon(\mathbf{x}) := \mathbf{K}(\mathbf{x}, \frac{\mathbf{x}}{\varepsilon})$
 195 where $\mathbf{K}(\mathbf{x}, \mathbf{y})$ is symmetric and continuous for all $(\mathbf{x}, \mathbf{y}) \in \Omega^\varepsilon \times Y$ and $\mathbf{K}(\mathbf{x}, \cdot)$
 196 is Y -periodic. Further, the constants $\beta, \lambda > 0$ exist such that

$$197 \quad \beta \|\psi\|^2 \leq \psi^\top \mathbf{K}(\mathbf{x}, \mathbf{y}) \psi \leq \lambda \|\psi\|^2, \quad \text{for all } \psi \in \mathbb{R}^d, \mathbf{x} \in \Omega^\varepsilon \text{ and } \mathbf{y} \in Y.$$

198 *Remark 3.1.* Assumptions (B1) and (B2) are made in order to use the periodic
 199 homogenization theory for developing the multi-scale approach (see [20]). On the
 200 other hand, the assumptions (A1) to (A4) are required to formulate Theorem 3.2 (see
 201 [4, 12, 30]).

202 **3.1. The homogenization approach.** A direct numerical approximation of the
 203 problem (2.3) requires the usage of an extremely fine mesh to capture all the changes

204 in the characteristics of the medium. In the following we consider an alternative
 205 approach and compute an effective model involving only the essential variations of
 206 the permeability matrix. Alternative approaches have also been considered, we refer
 207 to [36] for one involving the harmonic average of the parameters. Such techniques are
 208 rather suited for particular cases, e.g stratified media. However, our target is wider
 209 and the technique used here is mathematically consistent. We refer to [20, 26, 39] for
 210 a detailed presentation of the homogenization method.

211 Here, we restrict the presentation to the minimum needed for explaining the
 212 approach. We assume that all quantities satisfy the *homogenization ansatz* theory.
 213 For example p^ε can be formally expanded as power series in ε as

$$214 \quad (3.1) \quad p^\varepsilon(\mathbf{x}, t) = p_0(\mathbf{x}, \mathbf{y}, t) + \varepsilon p_1(\mathbf{x}, \mathbf{y}, t) + \varepsilon^2 p_2(\mathbf{x}, \mathbf{y}, t) + \dots$$

215 where $\mathbf{y} = \frac{\mathbf{x}}{\varepsilon}$ stands for the fast variable, \mathbf{x} is the slow variable and each function
 216 $p_i : \Omega^\varepsilon \times Y \times (0, T] \rightarrow \mathbb{R}$ is Y -periodic w.r.t \mathbf{y} . Additionally, the two-scale gradient
 217 and divergence operators become

$$218 \quad \nabla = \nabla_x + \frac{1}{\varepsilon} \nabla_y \quad \text{and} \quad \text{div} = \text{div}_x + \frac{1}{\varepsilon} \text{div}_y.$$

219 Using (3.1), the two-scale operators and since $b(\mathbf{x}, \mathbf{y}, p^\varepsilon) = b^\varepsilon(\mathbf{x}, p^\varepsilon)$ one applies the
 220 Taylor expansion of $b(\cdot, \cdot, p)$ about p_0 to obtain

$$222 \quad (3.2) \quad \partial_t b - \left(\text{div}_x + \frac{1}{\varepsilon} \text{div}_y \right) \left(\mathbf{K} \left(\nabla_x + \frac{1}{\varepsilon} \nabla_y \right) (p_0 + \varepsilon p_1 + \varepsilon^2 p_2) \right) + \mathcal{O}(\varepsilon) = f.$$

224 Equating similar terms in ε one gets that $p_0 = p_0(\mathbf{x}, t)$ does not depend on \mathbf{y} and is
 225 in fact the *macro-scale approximation* of the pressure $p^\varepsilon(\mathbf{x}, t)$.

226 To determine p_1 as a function of p_0 , for the terms of order $\mathcal{O}(\varepsilon^{-1})$ we can write
 227 $p_1(\mathbf{x}, \mathbf{y}, t) = \hat{p}_1(\mathbf{x}, t) + \sum_{j=1}^d \frac{\partial p_0(\mathbf{x}, t)}{\partial x_j} \omega^j(\mathbf{x}, \mathbf{y})$ where the function \hat{p}_1 is an arbitrary
 228 function of \mathbf{x} and ω^j are the Y -periodic solutions of the following *micro-cell* problems

$$229 \quad (3.3) \quad -\nabla_y \cdot (\mathbf{K}(\mathbf{x}, \mathbf{y}) (\nabla_y \omega^j + \mathbf{e}_j)) = 0, \quad \text{for all } \mathbf{y} \in Y.$$

230 Here $\{\mathbf{e}_j\}_{j=1}^d$ is the canonical basis of dimension d . To guarantee the uniqueness
 231 of the solution we assume that ω^j has the average 0 over the micro cells, that is,
 232 $\int_Y \omega^j(\mathbf{x}, \mathbf{y}) = 0$ for all $\mathbf{x} \in \Omega$.

233 To simplify the notation in the following we use p instead of p_0 for the *macro-scale*
 234 *approximation* of the pressure p^ε . Recalling the boundary conditions on the micro-
 235 scale, the terms of order zero in (3.2) and averaging over Y one obtains the following
 236 *homogenized problem*

$$237 \quad (3.4) \quad \begin{aligned} \partial_t b^*(\mathbf{x}, p) - \text{div}(\mathbf{K}^*(\mathbf{x}) \nabla p) &= f^*(\mathbf{x}, t), & \text{in } \Omega_T := \Omega \times (0, T], \\ p &= 0, & \text{on } \partial\Omega, \\ p(\mathbf{x}, 0) &= p_I, & \text{in } \Omega. \end{aligned}$$

238 The effective permeability $\mathbf{K}^* : \Omega \rightarrow \mathbb{R}^{d \times d}$ has the elements

$$239 \quad \mathbf{K}_{i,j}^*(\mathbf{x}) = \int_Y (\mathbf{K}(\mathbf{x}, \mathbf{y}) (\mathbf{e}_j + \nabla_y \omega^j(\mathbf{x}, \mathbf{y}))) \cdot \mathbf{e}_i \, d\mathbf{y}, \quad (i, j = 1, \dots, d).$$

240 The upscaled saturation and source are

$$241 \quad b^*(\mathbf{x}, p) := \int_Y b(\mathbf{x}, \mathbf{y}, p) \, d\mathbf{y} \quad \text{and} \quad f^*(\mathbf{x}, t) := \int_Y f(\mathbf{x}, \mathbf{y}, t) \, d\mathbf{y}.$$

242 The difference between the original problem (2.1) and the approximated problem
 243 (3.4) is subtle. In the original problem, the main characteristics are present at all
 244 scales, in the complex domain and in a strongly coupled manner. The *homogenized*
 245 *model* instead involves only essential variations at the macro-scale. However, to deter-
 246 mine the value of the permeability tensor at a macro point $\mathbf{x} \in \Omega$, one has to solve d
 247 micro-cell problems (3.3) associated with that macro point. Note that these problems
 248 reflect the rapidly oscillating characteristics and are decoupled from the macro-scale
 249 variations. From a computational point of view, the importance of this decoupling be-
 250 comes obvious. Instead of solving the full problem, one solves a collection of simpler
 251 problems. In general, analytic solutions are not available to compute the homoge-
 252 nized parameters. Then \mathbf{K}^* , b^* and f^* must usually be computed numerically and
 253 can therefore only be obtained at discrete points of the domain Ω . This strategy was
 254 also used in [1, 3, 5].

255 Concerning the existence and uniqueness of the weak solution of the homogenized
 256 problem (3.4) we use the assumptions (A1) to (A4) and assumptions (B1) and (B2)
 257 and refer to [4]. More precisely if there exists a constant $\theta > 0$ such that, for every δ
 258 and R with $0 < \delta < R$ there exists $C(\delta, R) > 0$ such that

$$259 \quad |b(\mathbf{x}, \mathbf{y}, \rho_1) - b(\mathbf{x}, \mathbf{y}, \rho_2)| \geq C(\delta, R) |\rho_1 - \rho_2|^\theta,$$

260 for all $\mathbf{x} \in \Omega^\varepsilon$, $\mathbf{y} \in Y$ and $\rho_1, \rho_2 \in [-R, R]$ with $\delta < |\rho_1|$, then the strong convergence
 261 of p^ε to p is showed in the following theorem (see [12, 27, 30] for the proof).

262 **THEOREM 3.2.** *Let p^ε be a family of solutions of the problem (2.1). If p^ε is such*
 263 *that $\sup_\varepsilon \|p^\varepsilon\|_{L^\infty(\Omega_T^\varepsilon)} \leq C$ with $C > 0$ and under the assumptions (A1)-(A4) and*
 264 *(B1)-(B2), there exists a subsequence of p^ε , still denoted by p^ε , such that for all q*
 265 *with $0 < q < \infty$, we have, $p^\varepsilon \rightarrow p$ strongly in $L^q(\Omega_T)$, where p solves (3.4).*

266 Following the ideas mentioned in Section 2, defining $\mathbf{u}(\mathbf{x})$ as the upscaled Darcy
 267 velocity, the upscaled unknowns $(p, \mathbf{u}) \in L^2(\Omega) \times H(\operatorname{div}, \Omega)$ satisfy

$$268 \quad (3.5) \quad \begin{aligned} \partial_t b^*(\mathbf{x}, p(\mathbf{x}, t)) + \operatorname{div}(\mathbf{u}(\mathbf{x}, t)) &= f^*(\mathbf{x}, t), & \text{in } \Omega_T, \\ \mathbf{u}(\mathbf{x}, t) &= -\mathbf{K}^*(\mathbf{x}) \nabla p(\mathbf{x}, t), & \text{in } \Omega_T, \\ p(\mathbf{x}, t) &= 0, & \text{on } \partial\Omega, \\ p(\mathbf{x}, 0) &= p_I, & \text{in } \Omega. \end{aligned}$$

269 *Remark 3.3.* If the original permeability \mathbf{K}^ε satisfies (B2) then the effective tensor
 270 \mathbf{K}^* is also symmetric and positive definite. Nevertheless, in the case of an initial
 271 isotropic medium the effective permeability can contains anisotropies, (e.g the tensor
 272 could be non-diagonal). However, in this case the non-diagonal components of \mathbf{K}^* are
 273 neglectable and the diagonal elements are similar.

274 The non-linear discrete problem associated with the homogenized formulation
 275 (3.5) is defined in the following.

276 **3.2. The non-linear fully discrete homogenized problem.** Let \mathfrak{T}_H be a
 277 coarse-triangular partition of the domain Ω with coarse elements \mathcal{T} of diameter $H_{\mathcal{T}}$
 278 and $H := \max_{\mathcal{T} \in \mathfrak{T}_H} H_{\mathcal{T}}$. For the discretization of the flux \mathbf{u} we consider the lowest-order
 279 Raviart-Thomas space $V_H := \mathcal{RT}_0(\mathfrak{T}_H)$ and for the pressure p we use the discrete
 280 subspace of piecewise constant functions W_H (see [10]).

281 **Problem (PH_n)**. For a given $(p_H^{n-1}, \mathbf{u}_H^{n-1}) \in W_H \times V_H$ and $n \geq 1$, find $p_H^n \in W_H$
 282 and $\mathbf{u}_H^n \in V_H$ such that for any $q_H \in W_H$ and $\mathbf{v}_H \in V_H$ there holds

$$283 \quad (3.6) \quad \begin{aligned} \langle b^*(\cdot, p_H^n) - b^*(\cdot, p_H^{n-1}), q_H \rangle + \Delta t \langle \operatorname{div}(\mathbf{u}_H^n), q_H \rangle &= \Delta t \langle f^*, q_H \rangle, \\ \langle [\mathbf{K}^*]^{-1} \mathbf{u}_H^n, \mathbf{v}_H \rangle - \langle p_H^n, \operatorname{div}(\mathbf{v}_H) \rangle &= 0. \end{aligned}$$

284 Again p_H^0 is the L^2 -projection of the initial p_I over the coarse mesh \mathfrak{T}_H . For simplicity
 285 we omit writing the \mathbf{x} argument in $b^*(\mathbf{x}, p)$, which becomes now $b^*(p)$. If $b(\mathbf{x}, \mathbf{y}, p_H^n)$
 286 has no micro-scale dependence $b^*(p_H^n) = |P| b(\mathbf{x}, p_H^n)$ and the same argument applies
 287 for f^* .

288 **3.3. Micro-cell problem and micro-scale discretization.** As mentioned be-
 289 fore, the effective parameters must be computed at each integration point on the coarse
 290 triangulation \mathfrak{T}_H . The effective tensor \mathbf{K}^* depends on the solution of the micro cell
 291 problems (3.3). To solve (3.3) we use the same MFEM scheme as for (3.4).

292 To approximate the solution of (3.3) we use a triangular decomposition \mathfrak{T}_h of the
 293 micro-scale domain $P \subseteq Y$ with micro-scale mesh size h . For each integration point
 294 $\mathbf{x} \in \mathcal{T}$ and $\mathcal{T} \in \mathfrak{T}_H$, the discrete micro-cell problem is

295 **Problem (Ph_j)**. Find $(\omega_h^j, \vec{\xi}_h^j) \in W_h \times V_h$ satisfying

$$296 \quad (3.7) \quad \begin{aligned} \langle \operatorname{div}(\vec{\xi}_h^j), q_h \rangle &= \langle f, q_h \rangle, \\ \langle [\mathbf{K}(\mathbf{x}, \cdot)]^{-1} \vec{\xi}_h^j, \mathbf{v}_h \rangle - \langle \omega_h^j, \operatorname{div}(\mathbf{v}_h) \rangle &= 0, \\ \omega_h^j &\text{ is } Y\text{-periodic,} \end{aligned}$$

297 for all $q_h \in W_h$, $\mathbf{v}_h \in V_h$ and $j = 1, \dots, d$.

298 After solving (3.7) we can compute the discrete effective permeability: $\mathbf{K}_{i,j}^*(\mathbf{x}) =$
 299 $\left(\int_Y \left(\mathbf{K}(\mathbf{x}, \mathbf{y}) \left(\mathbf{e}_j + \vec{\xi}_h^j(\mathbf{y}) \right) \right) \cdot \mathbf{e}_i \, d\mathbf{y} \right)$ and use it to solve the discrete problem (3.6).
 300 Note that these cell problems only need to be solved initially, or when the mesh
 301 changes.

302 **4. Adaptive numerical homogenization.** The standard homogenization the-
 303 ory applies for periodic media although its extension to random media is well un-
 304 derstood (see e.g [2]). In practical cases, one does not have any structure in the
 305 oscillations of the data. Nevertheless the computation of macro-scale parameters re-
 306 mains a suitable idea. We propose to solve the micro-cell problems (3.7) and compute
 307 the macro-scale parameters over a coarse mesh defined by the user. This procedure
 308 consists in two steps:

- 309 • *The macro-scale partition:* Define a macro-scale division of the domain Ω
 310 with elements Q_k , ($k = 1, 2, \dots, m$), where m is the total number of coarse
 311 cells.
- 312 • *The micro-scale domains:* Solve the micro-cell problems (3.7) over each coarse
 313 element Q_k . Note that Q_k determines a micro-scale domain and there we
 314 define a micro-scale mesh size h . The upscaled permeability that belongs to
 315 each region Q_k highly depends on the choice of h .

316 Subsequently one can mesh the macro-scale domain and solve the homogenized prob-
 317 lem (3.4). In Figure 2 we show the configuration of the macro and micro-scale partition
 318 and the procedure described previously. Note that neither the macro-scale partition
 319 nor the micro-scale mesh needs to be uniform.

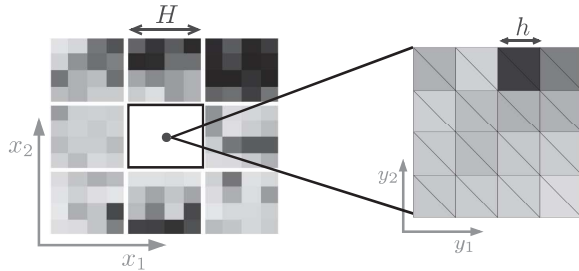


FIG. 2. Sketch of the macro-scale partition and the correspondent micro-scale discretization in a domain $\Omega \subset \mathbb{R}^2$. Different colours represent different values of the permeability.

320 **4.1. A-posteriori error estimates.** In the following we propose a four-step
 321 strategy to localize the error of the numerical solution of the homogenized problem.
 322 For this reason, it is necessary to compute a local error estimator $\eta_{\mathcal{T}}$ at each element
 323 $\mathcal{T} \in \mathfrak{T}_{H_n}$. Choosing an error estimator $\eta_{\mathcal{T}}^n$ highly depends on the features of each
 324 numerical method, the approximation, the post processing strategy, the implementa-
 325 tion etc. Nevertheless, we refer to [14], where the equivalence between different
 326 a-posteriori error estimators is analysed.

327 In the *error control based on averaging technique* [13, 15] the idea is to estimate
 328 the error based on a smoother approximation to the discrete solution \mathbf{u}_H^n . We define
 329 a global error estimator η_{Ω} and an average operator \mathfrak{A}_z

$$330 \quad \eta_{\Omega} := \min_{\mathbf{v} \in V_h} \|\mathbf{u}_H^n - \mathbf{v}\|_{L^2(\Omega)} \quad \text{and} \quad \mathfrak{A}\mathbf{u}_H^n(z) = \mathfrak{A}_z(\mathbf{u}_H^n) := \frac{1}{|w_z|} \int_{w_z} \mathbf{u}_H^n \, dx$$

331 where $w_z := \text{int}(\cup \{\mathcal{K} \in \mathfrak{T}_{H_n} : \mathcal{K} \cap \mathcal{T} \neq \emptyset, z \in \mathcal{T}\})$ is the patch corresponding to the
 332 point $z \in \Omega$. It has been proven (see [13]) that the error $\|\mathbf{u}^n - \mathbf{u}_H^n\|_{L^2(\Omega)}$ is bounded
 333 by $\|\mathbf{u}_H^n - \mathbf{v}\|_{L^2(\Omega)}$ for any continuous and piecewise polynomial \mathbf{v} . Then an upper
 334 bound of η_{Ω} can be computed as

$$335 \quad (4.1) \quad \|\mathbf{u}^n - \mathbf{u}_H^n\|_{L^2(\Omega)} \leq C\eta_{\Omega} + \text{h.o.t} \leq C\|\mathbf{u}_H^n - \mathfrak{A}\mathbf{u}_H^n\|_{L^2(\Omega)} + \text{h.o.t}$$

336 for some $C > 0$ independent of the mesh size. After choosing this estimator and in
 337 order to find the *optimal* macro-scale division to compute the effective parameters, as
 338 well as the solution of the homogenized problem (3.4) we will use a mesh adaptivity
 339 strategy.

340 **4.2. Mesh adaptivity.** We continue with a mesh adaptivity process using the
 341 a-posteriori estimator (4.1). Our approach consists of the sequence: Solve - estimate
 342 the error - select the cells/triangles - refine the mesh. The mesh refining generates a
 343 sequence of triangular meshes (one mesh per time step).

344 (S1) *Solve*: The starting point is an initial coarse mesh \mathfrak{T}_{H_1} and the approximation
 345 of the pressure and velocity (p_H^1, \mathbf{u}_H^1) that satisfy the discrete problem (3.6)
 346 in the first time step.

347 (S2) *Estimate the error*: Let the solution (p_H^n, \mathbf{u}_H^n) over \mathfrak{T}_{H_n} be given. Locally,
 348 an upper bound error estimator can be computed using the element-wise
 349 contributions in (4.1), i.e $(\eta_{\mathcal{T}}^n) := \|\mathbf{u}_H^n - \mathfrak{A}\mathbf{u}_H^n\|_{L^2(\mathcal{T})}$.

350 (S3) *Select the cells/triangles*: An *optimal* mesh corresponds to a mesh where the
 351 error is equidistributed. For this reason, the elements marked to be refined
 352 are $\mathcal{T} \in \mathfrak{T}_{H_n}$ such that (see [16])

$$353 \quad \eta_{\mathcal{T}}^n \geq \Theta \left(\max_{\mathcal{K} \in \mathfrak{T}_{H_n}} \eta_{\mathcal{K}}^n \right) \quad \text{with } \Theta \in (0, 1).$$

354 (S4) *Refine the mesh*: The last step of the refinement corresponds to including
 355 new points and re-mesh. Our strategy avoids the possibility of nonconforming
 356 meshes. We refine each selected cell in four new cells to compute four new
 357 the effective permeabilities. Inside of the new finer cells we re-mesh with the
 358 necessary triangles.

359 The outline of the steps (S1) to (S4) is presented in Figure 3 for the 2D case and
 360 in 3D the refinement can be done as described in [24]. In Figure 3 we highlight that
 361 at every time step it is necessary to make sure that in the new mesh each element
 362 corresponds only to one permeability value. That restriction forces us to refine also
 363 neighbouring elements and increases the resolution of the numerical solution.

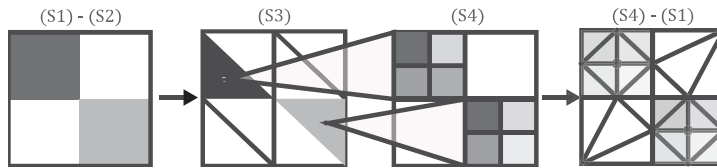


FIG. 3. Outline of the a -posteriori refinement in \mathbb{R}^2 . (Left to right) Initial effective permeability. Initial triangulation and selected triangles to refine. Refinement of the permeability field. Refinement of the triangular mesh.

364 With this strategy we allow to have more than one level of refinement, although
 365 the homogenization theory only consider two levels. The threshold for the refinement
 366 Θ can be chosen depending on the problem. We remark that higher values of Θ remain
 367 into coarser meshes and less error control. In the numerical examples we show how
 368 the refinement increase the accuracy of the parameters and the upscaled solution.

369 **5. Linearization.** A popular strategy to solve non-linear problems is Newton's
 370 method (see [8]). The reason to use Newton's method is the quadratic convergence,
 371 but we remark that quadratic convergence only arises under certain restrictions and it
 372 is only locally convergent. For the Newton method, the initial guess for the iterations
 373 must to be close enough to the expected solution for the scheme to be convergent.
 374 For all these reasons, we will also use a fixed point iteration scheme, called L-scheme.
 375 Although the L-scheme is only linearly convergent, it has unconditional convergence,
 376 meaning that it converges to the time-discrete solution regardless of the initial point
 377 and it does not involve any derivatives (see [34, 31, 38]).

378 For a $\mathfrak{L} \geq \max_{p \in \mathbb{R}} \{\partial_p b^*(\cdot, p)\}$, assume p_H^{n-1} given. With $i \in \mathbb{N}$, $i \geq 1$ being
 379 the iteration index, the L-Scheme is introduced through: Find $p_H^{n,(i)} \in W_H$ and
 380 $\mathbf{u}_H^{n,(i)} \in V_H$ such that for any $q_H \in W_H$ and $\mathbf{v}_H \in V_H$ there holds

$$\begin{aligned}
 & \left\langle \mathfrak{L} \left(p_H^{n,(i)} - p_H^{n,(i-1)} \right) + b^* \left(\cdot, p_H^{n,(i-1)} \right), q_H \right\rangle \\
 381 \quad (5.1) \quad & + \Delta t \left\langle \operatorname{div} \left(\mathbf{u}_H^{n,(i)} \right), q_H \right\rangle = \Delta t \left\langle f^*, q_H \right\rangle + \left\langle b^* \left(\cdot, p_H^{n-1} \right), q_H \right\rangle, \\
 & \left\langle \mathbf{u}_H^{n,(i)}, \mathbf{v}_H \right\rangle - \left\langle \mathbf{K}^* p_H^{n,(i)}, \operatorname{div} \left(\mathbf{v}_H \right) \right\rangle = 0.
 \end{aligned}$$

382 Where the natural choice for the initial iteration $p_H^{n,0}$ is p_H^{n-1} . In our non-linear solver
 383 the iterations take place until one reaches a prescribed threshold for the L^2 -norm of
 384 the residual $\partial p_H^{n,i} := p_H^{n,(i)} - p_H^{n,(i-1)}$.

385 The use of an upper bound of $\partial_p b^*(\cdot, p)$ affects the convergence rate. For the
 386 L-scheme the convergence rate is $\alpha = \frac{\mathfrak{L}-m}{\mathfrak{L}+C\Delta t}$ for some $C > 0$ and $m < \mathfrak{L}$ (see [31]).

387 This leads to an extremely slow convergences in some cases (e.g large \mathfrak{L} or small Δt).
 388 For this reason in Section 6 we choose a smaller value $\mathfrak{L} = \frac{1}{2} \max_{p \in \mathbb{R}} \{\partial_p b^*(\cdot, p)\}$ which
 389 still gives convergence (see [28]). For more results and analysis of the linearization
 390 techniques we refer to [31, 28] and therein references.

391 **6. Numerical results.** We present two numerical examples in \mathbb{R}^2 to illustrate
 392 the behaviour of the proposed adaptive homogenization procedure. We first verify
 393 our numerical homogenization approach using a manufactured periodic and quasi-
 394 periodic media and subsequently use a non-periodic test case. Note that all parameters
 395 specified in the following examples are non-dimensional and the pressures are also
 396 shifted to lie between 0 and 1.

397 **6.1. Periodic and quasi-periodic cases.** Consider the macro-scale domain
 398 $\Omega = [0, 1] \times [0, \frac{1}{2}]$ with initial condition $p_0 = 0$ and no-flux boundary conditions. The
 399 isotropic periodic permeability field is defined by

$$400 \quad \mathbf{K}^\varepsilon(\mathbf{x}) = \left(10x_1^2 x_2 + \frac{1}{2 + 1.8 \cos(2\pi \frac{x_1}{\varepsilon}) \cos(2\pi \frac{x_2}{\varepsilon})} \right) \mathbb{I}_{2 \times 2}$$

401 A source and a sink are placed in the upper-right and the lower-left corners, having
 402 fixed pressures of 1 and 0, respectively. The volumetric concentration is $b^\varepsilon(\mathbf{x}, p) =$
 403 $\mathcal{R}(p^\varepsilon)^3$. Here \mathcal{R} is a non-dimensional constant that let us simulate a *fast diffusion*
 404 process. For the time discretization we take $T = 1$ with $\Delta t = 0.02$.

405 To solve the problem (5.1) with the necessary resolution to capture the oscillations
 406 over Ω the mesh size is restricted to be $h^\varepsilon \ll \varepsilon$. We use $h^\varepsilon = 5 \times 10^{-3}$ to compute
 407 the fine-scale solutions $(p_{h^\varepsilon}, \mathbf{u}_{h^\varepsilon})$ when $\varepsilon = \frac{1}{8}, \frac{1}{16}$ and $\frac{1}{32}$. The reference solutions
 408 are computed using the same MFEM, backward Euler scheme and the L-scheme with
 409 $\mathfrak{L} = 1.5 \frac{\mathcal{R}}{2} \geq \frac{\max(3\mathcal{R}(p^\varepsilon)^2)}{2}$.

410 Table 1 shows the history of convergence of the error for different values of ε and
 411 three coarse meshes \mathfrak{T}_H without refinement and $H \gg h^\varepsilon$. The relative L^2 -error e_H
 412 in Table 1 is $e_H = \frac{\|\Pi_{h^\varepsilon}(p_H) - p_h\|_{L^2(\mathfrak{T}_{h^\varepsilon})}}{\|p_h\|_{L^2(\mathfrak{T}_{h^\varepsilon})}}$ where $\Pi_{h^\varepsilon}(p_H)$ is the projection
 413 of the coarse-scale solution in the fine mesh $\mathfrak{T}_{h^\varepsilon}$. With this result we show that the
 414 homogenized solution converges to the solution of the original problem when $H \rightarrow 0$
 415 and also when $\varepsilon \rightarrow 0$.

416 Nevertheless, in the following we use a modified permeability field to ensure that
 417 any assumption of periodicity is necessary. We include in the same domain Ω a high
 418 permeability region Ω_1 and a low permeability region Ω_2 where the permeability is
 419 10^{-2} and 10^{-7} respectively.

$$420 \quad \Omega_1 := [0.21, 0.41] \times [0.11, 0.41] \text{ and } \Omega_2 := \{\mathbf{x} \in \Omega^\varepsilon \mid \|\mathbf{x} - [0.75, 0.26]\|_2 \leq 0.1^2\}.$$

421 In Figure 4 the normalized (quasi-periodic) permeability field is showed for two
 422 values of the scale parameter ε . In this case the boundary conditions, the volumetric
 423 concentration, the source term and the time discretization remain the same as before.
 424 Figure 5 shows four levels of the first component of the effective permeability tensor
 425 $(\mathbf{K}_{1,1}^*)$ starting with a coarse grid of 16×8 cells. Referring to the different levels of
 426 the effective permeabilities is important to remark that the coarse-scale permeabilities
 427 are computed in zones that not always match with the initial resolution or periodicity.
 428 Here one can notice the influence of neighbouring macro-cells in the numerical
 429 solution of the micro problems (3.7). This effect is evident at the boundary of the
 430 low permeability zone Ω_2 . To point out this behaviour in the Figure 5 we highlight

ε	H	Relative error (e_H)
1/8	0.1768	0.1938
1/8	0.0884	0.1287
1/8	0.0442	0.0856
1/16	0.1768	0.1797
1/16	0.0884	0.1138
1/16	0.0442	0.0724
1/32	0.1768	0.1690
1/32	0.0884	0.1030
1/32	0.0442	0.0621

TABLE 1

History of convergence of the error for three values of ε and three coarse meshes.

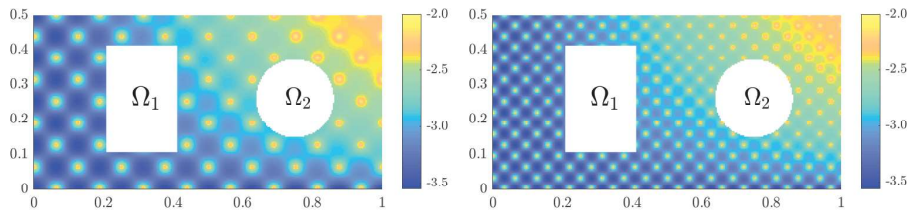


FIG. 4. Fine scale permeability field (left) $\varepsilon = \frac{1}{8}$ and (right) $\varepsilon = \frac{1}{16}$ (Log_{10} scale).

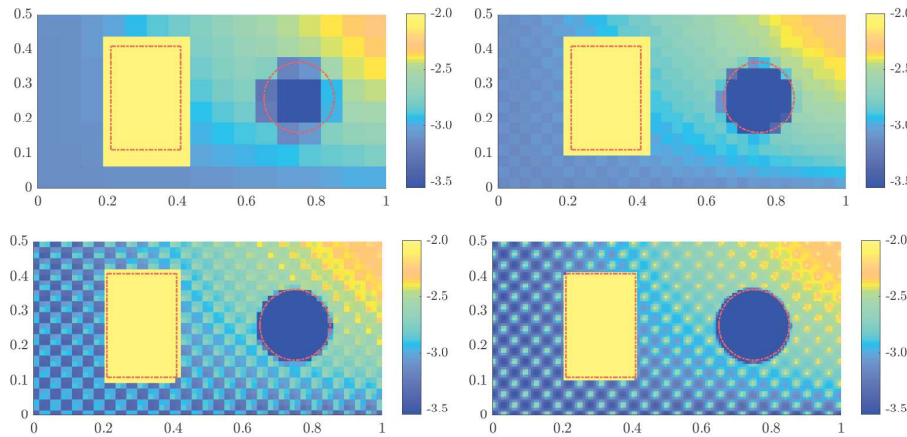


FIG. 5. Coarse-scale permeability distribution (Log_{10} scale) starting with a coarse grid of 16×8 cells. The red lines indicates the original location of the low permeability zone ($\mathbf{K}^\varepsilon = 10^{-7}$) and high permeability zone ($\mathbf{K}^\varepsilon = 10^{-2}$).

431 with a dashed lines the original location of the low and high permeability areas. The
 432 numerical solution of the lineal upscaled problem (5.1) is showed in Figure 6. The

433 upscaled solution is computed using the mesh adaptivity described in Section 4 using
 434 the threshold for the mesh adaptivity $\Theta = 0.5$. At the end of the adaptive process,
 435 the relative L^2 -error of the upscaled pressure p_{H_n} is 1.6% using only the 14.7% of the
 436 original degrees of freedom.

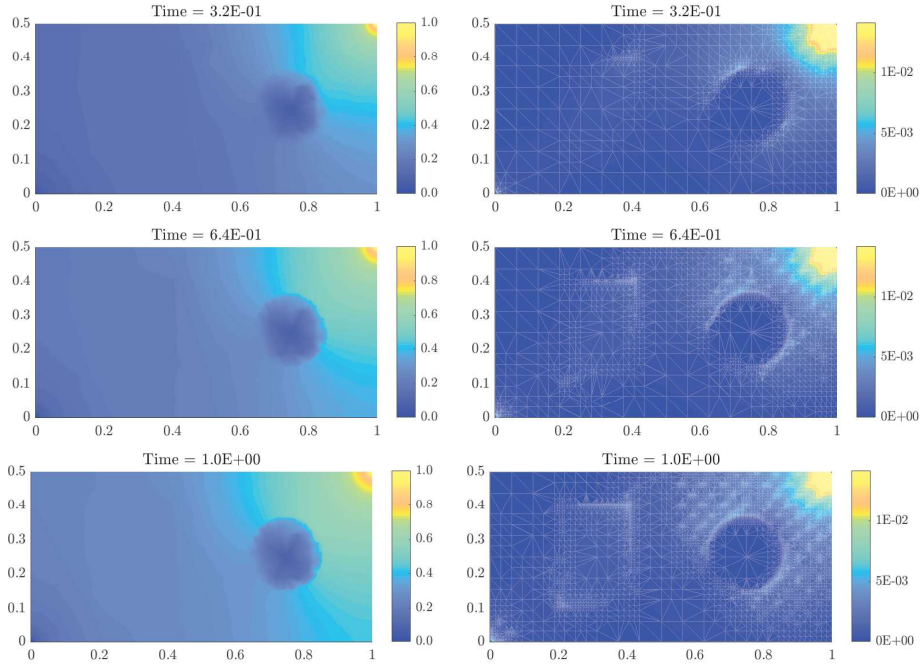


FIG. 6. Adaptive homogenization at $t = 16\Delta t$, $32\Delta t$, $50\Delta t$. Pressure p_{H_n} (left) and magnitude of the velocity field $\|u_{H_n}\|_2$ (right) over meshes with 2.367, 5.950 and 9.659 coarse elements.

437 Furthermore, after the adaptivity process we obtain a refined version of the per-
 438 meability field and Figure 7 shows the result of the refined permeability at $t = 1$.

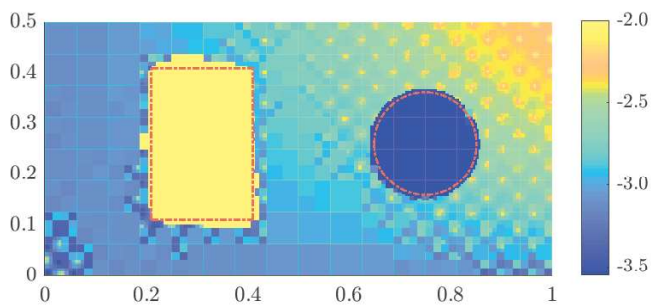


FIG. 7. Refined permeability field at $t = 1$ (Log_{10} scale).

439
 440 Concerning the behaviour of the non-linear solver, our test case is an example
 441 where the convergence of the Newton's method highly depends of the initial guess.
 442 However the convergence of the L-scheme is not optimal; i.e., even though the L-
 443 scheme converges we do not want to lose the quadratic convergence of the Newton's

444 method. To compute the solution of the Figure 6 the linear solver using only the
 445 L-scheme reaches the threshold $\|\delta(p_H)\|_2 < 10^{-10}$ after an average of 70 iterations.
 446 In order to improve the linear solver we use a mixed strategy (see [9, 29]). The target
 447 is to construct an initial solution that suits a non-problematic starting point for the
 448 Newton's method. In this case we used the L-scheme until $\|\delta(p_H)\|_2 < 10^{-2}$ and then
 the classic Newton's method until one reaches $\|\delta(p_H)\|_2 < 10^{-10}$ (see Figure 8).

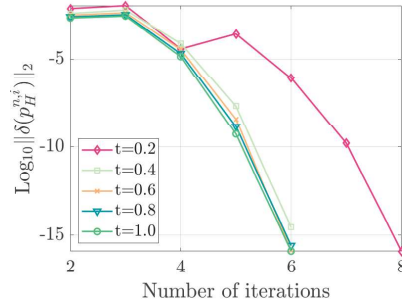


FIG. 8. Convergence of the residual in the non-linear solver. Results for four different times steps using the L-scheme with $\mathfrak{L} = 1.5 \frac{\mathcal{R}}{2}$ and Newton's method afterwards.

449

450 **6.2. Non-periodic case.** Here we consider a highly heterogeneous and non-
 451 periodic medium. We utilize the data of the SPE Comparative Solution Projects [19].
 452 This provides a vehicle for independent comparison of methods and a recognized suite
 453 of test datasets for specific problems. Our isotropic permeability field \mathbf{K}^ε is defined
 454 by the top field of SPE10th data set (see Figure 9).

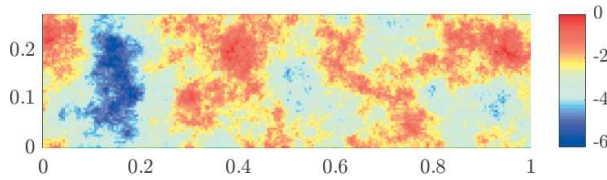


FIG. 9. Fine scale permeability distribution for SPE10th-TopLayer (Log_{10} scale).

455 The macro-scale domain is a two-dimensional rectangle (see Figure 9). External
 456 boundaries are impermeable; i.e., we take no-flux boundary conditions. The domain
 457 is initialized with pressure $p_0 = 0$. A source and a sink are placed in the lower-left
 458 and the upper-right corners, having fixed pressures of 1 and 0, respectively.

459 Moreover, the volumetric concentration is $b^\varepsilon(\mathbf{x}, (p^\varepsilon)) = \mathcal{R}(p^\varepsilon)^3$. Here \mathcal{R} is defined
 460 as in subsection 6.1. For the time discretization we take $T = 1$ with $\Delta t = 0.02$ and
 461 the parameter for the non-linear solver is $\mathfrak{L} = 1.5 \frac{\mathcal{R}}{2} \geq \frac{\max(3\mathcal{R}(p^\varepsilon)^2)}{2}$.

462 The adaptivity criteria for the dynamic mesh refinement, described in Section 4,
 463 is $\Theta = 0.2$. In this case we choose a value of Θ smaller than in subsection 6.1 because
 464 we address to capture more changes in the flux and those changes are related with
 465 the heterogeneity of the medium. To solve the problem (5.1) with the resolution of
 466 Figure 9 we construct a grid with 26.400 elements in a homogeneous triangular mesh
 467 $(\mathfrak{T}_{h^\varepsilon})$. In Figure 10 we show the reference solution $(p_{h^\varepsilon}, \mathbf{u}_{h^\varepsilon})$.

468 Using a coarse grid of 55×15 squares where we compute the first effective perme-
 469 ability field. This coarse grid corresponds to a macro-scale mesh with 1.650 triangular

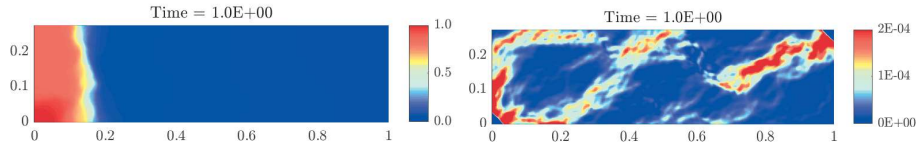


FIG. 10. Fine scale pressure p_h (left) and (right) magnitude of the velocity field $\|\mathbf{u}_h\|_2$.

470 elements. In Figure 11 we show the first component ($\mathbf{K}_{1,1}^*$) of the coarse-scale permeability field and this distribution is used afterwards to compute the first step of the
 471 adaptivity procedure.

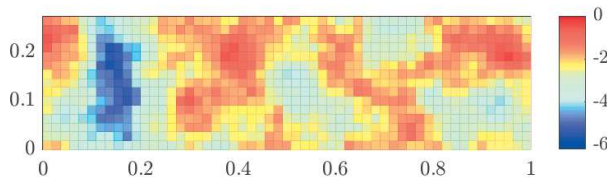


FIG. 11. Coarse-scale permeability distribution (Log_{10} scale).

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In Figure 12 we show the difference between the effective permeabilities computed
 474 with homogenization and using the harmonic average. The difference between these
 475 strategies is higher in zones with high permeability and one can point out that the
 476 harmonic average always underestimate the permeability. This is problematic because
 477 the high permeability regions are regions where one should increase the accuracy of
 478 the effective parameter in order to have better numerical solutions. When we compute
 479 the numerical solution of (5.1) using the harmonic average of the permeability the
 relative L^2 -error of the pressure is 12.3%.

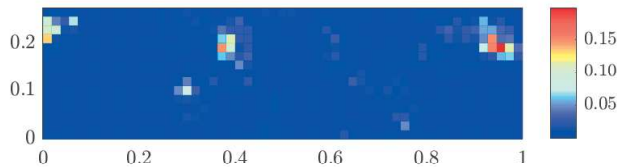


FIG. 12. Difference between the coarse-scale effective permeabilities using homogenization vs harmonic average.

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Figure 13 shows the numerical solution of the upscaled problem (5.1) using the
 482 mesh adaptivity described in section 4. At the end of the adaptive process, the relative
 483 L^2 -error of the upscaled pressure p_H is 4.72% using only the 16.5% of the original
 484 degrees of freedom. Furthermore, using the adaptivity process we obtain a refined
 485 version of the permeability field. Figure 14 shows the result of the permeability field
 486 after the refinement process.

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Finally, in Figure 15 we show the convergence of the norm of the residual $\delta(p_H)$
 when one use a combination of the L-scheme and Newton's method. Here we
 use a mixed strategy (see [29]) to construct an initial solution that suits a non-
 problematic starting point for the Newton's method. In this case we use the L-
 scheme until $\|\delta(p_H)\|_2 < 10^{-2}$ and then the classic Newton's method until one reaches
 $\|\delta(p_H)\|_2 < 10^{-10}$ and as we see in Figure 15 the quadratic convergence of the newton's
 method is recovered.

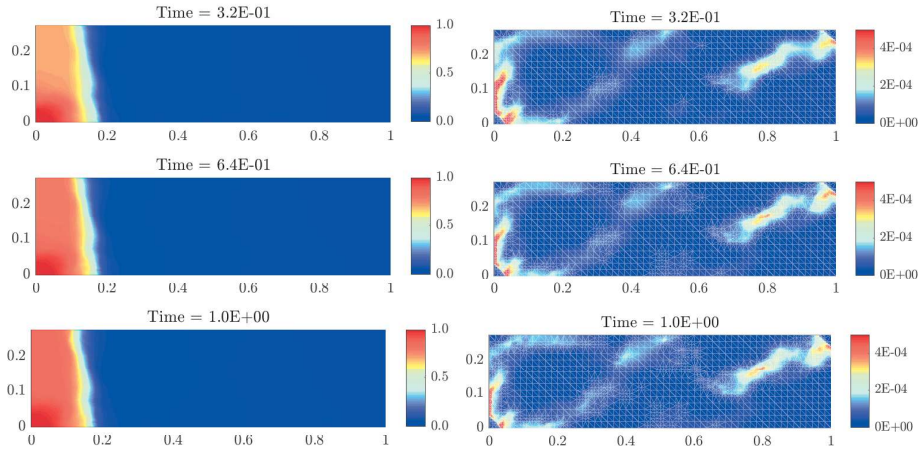


FIG. 13. Adaptive homogenization at $t = 16\Delta t$, $32\Delta t$, $50\Delta t$. Pressure p_{H_n} (left) and (right) magnitude of the velocity field $\|\mathbf{u}_{H_n}\|_2$ over meshes with 2.701, 3.573 and 4.353 coarse elements.

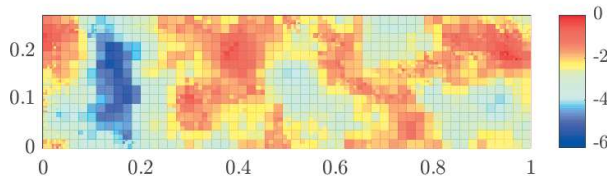


FIG. 14. Refined permeability field at $t = 1$ (Log_{10} scale).

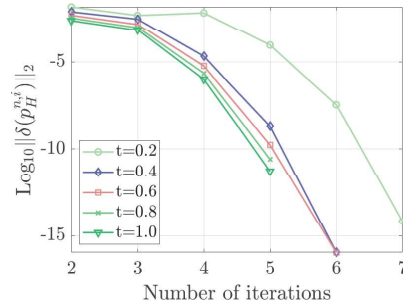


FIG. 15. Convergence of the residual in the non-linear solver. Results for five different times steps using the L -scheme with $\mathcal{L} = 1.5 \frac{\mathcal{R}}{2}$ and Newton's method afterwards.

494 **7. Conclusions.** We have proposed a numerical scheme based on homogeniza-
 495 tion to solve a non-linear parabolic equation with highly oscillatory characteristics.
 496 The discrete non-linear system is obtained by a backward Euler and lowest order
 497 Raviart-Thomas mixed finite element discretization. Our approach utilizes a local
 498 mesh refinement that leads to the computation of the effective parameters locally
 499 through decoupled cell problems. With this we achieved to improve the accuracy of
 500 the solution without compromising the efficiency of the method. The adaptivity is
 501 based on the idea that the upscaled parameters are updated only when it is necessary.
 502 Moreover, to illustrate the performance we have presented two general examples. We
 503 construct a periodic case to show the history of convergence of the error when the

504 scale separation tends to zero. In the non-periodic case we used a benchmark from
 505 the SPE10th project and we showed that the homogenization can be used in general
 506 cases.

507 In addition to the aforementioned, we combined the standard Newton's method
 508 and the L-scheme to improve the behaviour of the non-linear solvers. We presented a
 509 combination of techniques that lead to a very efficient numerical scheme. It is relevant
 510 to mention that besides the theory mentioned in this paper the applicability of this
 511 strategy is vast. Extensions of our adaptive algorithm including more complex micro-
 512 scale models are applicable. Those include from reactive transport up to moving
 513 interfaces affecting the structure of the micro-scale.

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