

## Responses to the comments

*We would like to express our gratitude and deep thanks to the reviewers and editors for taking the time to review our work and for their comments which help focus the goal of this research.*

I read the paper with interest. It presents a two-phase (oil and acidizing aqueous solution) thermal Darcy-Brinkman-Forchheimer (DBF) model for matrix acidization, incorporating temperature-dependent reaction kinetics. The related code is made available by the authors, which I did not try to compile and run myself. Key findings are claimed by the authors based on just seven 40x40 computed models with different initial or boundary conditions: the temperature of acid injection significantly affects dissolution patterns, while initial matrix temperature has minimal influence, and the imposed injection velocity plays also a role in shaping the dissolution pattern of the matrix.

Overall, the approach appears to be formally sound; however, in the present state the paper is seriously lacking some context as well as numerical applicability, and I cannot recommend it for publication without major-major revisions and integrations. I therefore recommend rejection prior to re-submission.

- The formulation appears sound and it is adequately described too - in the sense that the assumptions and simplifications are clear and evident for the reader familiar with both reactive transport and with multiphase flow. Philosophically, however, I have to admit that I would never advise to pursue a "global implicit" ADR model in this context, because no matter how complicated the model is made, it will be a very rough oversimplification in terms of chemistry and reactive transport, which are the controlling physical factors here.

Response: We thank the reviewer for the positive assessment of the formulation and for the insightful philosophical comment. We fully agree that, from a fundamental standpoint, chemical reactions and reactive transport processes are the dominant physical mechanisms in this problem, and that a global implicit ADR model inevitably represents a strong simplification of the underlying chemistry.

The objective of the present work, however, is not to provide a high-fidelity description of the detailed chemical kinetics or pore-scale reactive transport. Instead, our goal is to develop a computationally efficient model that captures the dominant macroscopic

flow–reaction coupling and the resulting wormhole evolution trends under a limited set of assumptions.

Within this context, the global implicit ADR formulation is intentionally adopted as a reduced-order representation, allowing us to focus on the interaction between flow, transport, and reaction at the continuum scale, while keeping the model tractable for analysis. We acknowledge that this approach is not intended to replace more detailed geochemical or reactive transport models, but rather to complement them in regimes where computational efficiency and qualitative predictive capability are of primary interest.

- Nothing is said about the implied formation matrix dissolved by acid injection. Is it a pure carbonate reservoir intended? What would change if there would be an inert matrix fraction, resisting to the acidification?

Response: We thank the reviewer for raising this important point. In the present work, the solid matrix is implicitly assumed to be a reactive carbonate matrix, representative of calcite-dominated carbonate formations commonly targeted by matrix acidizing treatments. This assumption is consistent with classical wormholing studies and continuum-scale acidizing models, where dissolution is dominated by acid–carbonate reactions.

We acknowledge that real reservoir rocks may contain inert or weakly reactive mineral fractions (e.g., quartz or clays), which would locally resist dissolution. Incorporating an inert fraction would primarily affect the porosity-evolution equation by introducing a reduced effective reactive surface area and a maximum attainable porosity. In practical terms, this would (i) slow down porosity growth, (ii) increase PVBT, and (iii) potentially suppress wormhole localization, especially at lower Damköhler numbers. To clarify this assumption, we will explicitly state in the revised manuscript that the model currently represents a single-reactive-mineral carbonate matrix.

- L 261: "rectangular matrix measuring 0.1 m × 0.1 m, discretized into a uniform grid of 40 × 40 cells": why has this particular length scale of 4 m been chosen? Anyhow, these numerical models can be taken as proof of concept for the validity of the formulation and of the numerical solution, but I am not convinced that such grid can be useful to depict real systems or draw meaningful conclusions about sensitivity of the process and the independence of the results from initial assumptions. Also, the images appear to be smoothly interpolated, hence misleading: a 40x40 image should be clearly "pixelated".

Response: We appreciate the reviewer's concern regarding domain size, grid resolution, and visualization. The selected domain size (0.1 m × 0.1 m) and grid resolution (40 × 40) are not intended to represent a field-scale reservoir, but rather to serve as a controlled numerical testbed for investigating the qualitative influence of thermal and hydrodynamic parameters on dissolution regimes in a two-phase thermal DBF framework.

This scale is consistent with many Darcy-scale proof-of-concept wormholing studies, where the objective is to resolve dissolution morphology rather than to achieve direct field-scale predictability. We fully agree that quantitative extrapolation to field conditions requires larger domains and higher resolutions. Thus, we have increased the grid size to 120 × 120, and redone all of the experiments.

Regarding visualization, we acknowledge that interpolation may visually obscure the underlying grid structure. However, it is the normal operation of many works on matrix acidizing. In the new manuscript, since the grid size is increased to 120 × 120, the negative effect of interpolation is greatly decreased.

- Line 266: "Porosity is spatially randomized with an average value of 0.18": how? Is it uniform white noise, or gaussian? Which range, which standard deviation? What happens if porosity is considered initially homogeneous?

Response: In fact, we use the function "random\_number (rand)" in FORTRAN90 to generate an array of pseudorandom numbers from the uniform distribution over the range [0, 1]. Then, we map the numbers in [0, 1] to another range [0.15, 0.21] by the formulation:  $\text{new\_rand} = 0.06 * \text{rand} + 0.15$ . The average value of new\_rand in [0.15, 0.21] is 0.18.

If porosity is considered initially homogeneous, a lot of dissolution patterns such as wormholes cannot be formed, since they depend on the heterogeneous attribute of the matrix.

- Figure 1: I do not see any appreciable difference in the three pictures. I believe it is necessary to find some other way to summarize the results, highlighting the negligible effect of temperature. Maybe just picture their differences?
- Figure 2-6: Same for Figure 1. At this point, figures 3 to 6 are redundant and the whole 4.1 section can be reduced to 1 figure for the lowest and highest temperatures, a mere proof of the negligible influence of the initial matrix temperature.

Response: We fully agree with the reviewer's assessment. The intent of Section 4.1 is precisely to demonstrate the negligible influence of initial matrix temperature, and we acknowledge that the current presentation is unnecessarily verbose.

In the revised manuscript, we have reduced Section 4.1 substantially, such as deleting many figures and retaining only two representative figures. Besides that, to make readers see the differences by eyeballs, we have added a new figure (Figure 3).

- Figure 7: things start to be interesting, some effects are visible, but again only qualitatively and not quantitatively. However, from here on basically all color scales for the juxtaposed images are not comparable anymore, which is not good scientific practice.

Response: We thank the reviewer for this important observation. We agree that consistent color scales are essential for meaningful comparison.

In the revised manuscript, all comparative figures except Figure 9 are: (1) Using identical color scales across cases, and (2) Clearly stating scale limits in captions. It is noted that the temperature scales are different for the two cases in Figure 9. If a unified temperature scale is given, the change of the temperature cannot be identified clearly

- Overall, sections 4.2 and 4.3 stretch the results of two and three different model runs respectively, without any quantification which can be assumed to be transferable (thing such as scale invariance, transferability of results to higher temperature...) or usable otherwise. Moreover, most figures do not actually add anything meaningful to the paper. This is the case for example for figures 10 and 11 (only one among 4 would suffice), but also figures 12 and 14, are they not just the expected consequences of the imposed boundaries?

Response: We thank the reviewer for this critical and thoughtful comment. We agree that Sections 4.2 and 4.3 are based on a limited number of representative simulations and that the results should not be interpreted as providing scale-invariant or universally transferable quantitative laws. This was not the intention of these sections. Rather, their primary purpose is to demonstrate and illustrate qualitative trends that emerge from the proposed two-phase thermal DBF framework under controlled conditions, and to verify that the model reproduces physically consistent behavior when key thermal and hydrodynamic parameters are varied.

Regarding quantification and transferability, we acknowledge that a systematic parametric or dimensionless analysis (e.g., in terms of Péclet or Damköhler numbers) would be required to establish broader generality. However, such an analysis would

substantially expand the scope of the manuscript. In the present work, we therefore deliberately restrict the conclusions to comparative trends within the tested parameter ranges, such as the observed analogy between increasing injection temperature and decreasing injection velocity.

Concerning the figures, we agree that some of them (e.g., Figures 10 and 11, and the temperature fields in Figures 12 and 14) largely reflect expected consequences of the imposed flow and thermal boundary conditions. Their inclusion is not intended to introduce new physical insights on their own, but rather to confirm the internal consistency of the coupled model, demonstrating that flow, saturation, dissolution, and temperature fields evolve coherently within the same simulation framework. In particular, the streamline and temperature plots help verify that the observed dissolution patterns are dynamically supported by the underlying flow and thermal fields, rather than being numerical artifacts.

We acknowledge that fewer figures could convey the same message more concisely, and we will consider minor reductions or clarifications in the figure presentation. Nevertheless, we believe that the current set of figures provides a complete and transparent illustration of the coupled thermo-hydro-chemical processes resolved by the model, which may be helpful for readers seeking to understand or reproduce the simulations.

- In my opinion, most of chapter 4 can be removed, hence highlighting the poorness of the presented results.
- Generally speaking, the authors completely neglect the abundant literature about the use of Péclet and Damkohler numbers to describe these class of models and the interacting boundary and initial conditions. This is scientifically unacceptable in my opinion. Properly familiarizing with this well known literature - a good starting point would be the Golfier paper, cited by the authors, and following other works referencing it - will also inevitably lead to more convincing design of experiment for the paper. As the manuscript currently is, these are merely nice pictures without real usability.

Response: We thank the reviewer for emphasizing the importance of dimensionless analysis and for pointing to the classical literature on dissolution regimes, including the work by Golfier et al., which we cite and highly value. We respectfully note, however, that the absence of an explicit Péclet–Damköhler number analysis does not imply a lack of scientific validity of the present study, nor is it uncommon in the matrix acidization literature.

Indeed, while dimensionless numbers provide a powerful framework for regime classification and scaling analysis, a substantial body of well-established work on matrix acidization—both Darcy-scale and Darcy–Brinkman–Forchheimer-based—investigates dissolution behavior primarily through controlled variations of physical parameters, most notably injection velocity, reaction rate constants, and transport coefficients, without explicitly reformulating the problem in terms of Péclet and Damköhler numbers. In practical acidizing studies, injection velocity is itself a primary operational control parameter and is widely used as a direct means to explore transitions between dissolution regimes. From a modeling standpoint, varying injection velocity effectively alters the balance between advection, diffusion, and reaction, and thus implicitly spans different Péclet and Damköhler number ranges.

The objective of the present work is not to re-establish dissolution regime diagrams or to perform a comprehensive scaling analysis, but rather to extend an existing DBF-based acidizing framework to a two-phase thermal setting and to examine, in a controlled and physically transparent way, how thermal effects interact with flow and reaction under typical acidizing conditions. Within this context, the numerical experiments are designed to highlight comparative trends—such as the analogous influence of injected acid temperature and injection velocity on dissolution patterns—rather than to claim universal or scale-invariant quantitative laws.

We agree that dimensionless analysis can be valuable for broader generalization, and we will clarify in the manuscript that the presented results are intended as mechanistic and model-demonstration studies, not as exhaustive regime maps. Nevertheless, we respectfully disagree that the current results are “merely nice pictures without usability.” On the contrary, they serve to verify that the proposed two-phase thermal DBF model reproduces physically consistent and literature-aligned behavior when key operational parameters are varied, thereby providing a sound basis for future, more extensive parametric or nondimensional studies.

- The Kozeny-Carman relationships should be probably somehow bounded for very small porosity values.

Response: We thank the reviewer for this remark regarding the applicability of the Kozeny–Carman relationship at extreme porosity values. We agree that, from a strictly theoretical standpoint, the classical Kozeny–Carman formulation may lose physical validity as porosity approaches its limiting values. However, we would like to emphasize that the use of Kozeny–Carman-type relationships is standard practice in matrix acidization modeling and has been widely adopted in both Darcy-scale and Darcy–Brinkman–Forchheimer-based studies of wormhole formation.

In the context of the present work, the Kozeny–Carman relationship is applied within a moderate and physically relevant porosity range, consistent with typical carbonate matrix acidization scenarios. The initial porosity is 0.18, and during the simulations the porosity evolves smoothly without approaching unrealistically small values. As a result, the numerical solution never enters a regime where porosity is close to zero and where additional bounding would become necessary to ensure stability or physical realism.

We note that many published acidizing studies employ the same permeability–porosity relationship without explicitly introducing lower or upper bounds, particularly when the focus is on dissolution-driven porosity increase rather than compaction or pore collapse. Within this commonly accepted modeling framework, the Kozeny–Carman relationship provides a consistent and physically interpretable link between porosity evolution and permeability enhancement.

- I can go on and list further deficiencies of the present manuscript. Just to name a few which I would have expected:
  1. lack of true control simulations, e.g. with initial homogeneous porosity;
  2. lack of code efficiency evaluation, things such as required CPU-time for the seven simulation; scaling of the computations with more CPUs (the code is claimed to be parallel) and so on.

These aspects are in my opinion fundamental for the GMD journal and must be properly addressed for consideration

Response: Regarding control simulations with homogeneous initial porosity, we note that the use of mildly heterogeneous initial porosity fields is a common and deliberate choice in wormholing simulations, as perfectly homogeneous conditions are known to suppress symmetry breaking and can lead to unrealistically uniform dissolution patterns. In this sense, the chosen setup is intended to reflect typical modeling practice rather than to explore all possible initial conditions. We agree that homogeneous-porosity cases can be useful as reference tests, but including such simulations would primarily serve a confirmatory role and would not alter the qualitative conclusions drawn in this work.

Concerning code efficiency and parallel scalability, we acknowledge that these topics are of interest to the GMD community. The present manuscript, however, builds upon a numerical framework whose parallel implementation, solver performance, and scalability characteristics have already been documented and validated in our previous publications. In this paper, we therefore focus on the model formulation and physical extensions, rather than repeating performance benchmarks that are largely unchanged by the addition of thermal and two-phase terms. For the relatively small number of

simulations presented here, computational cost is modest and does not constitute a limiting factor for the conclusions.

- I stress the fact that the posed model - in the sense of equations and possibly their numerical implementation - could really be interesting and worth of publication, but at the moment I see no convincing results.

# Two-Phase Thermal Simulation of Matrix Acidization Using the Non-Isothermal Darcy–Brinkman– Forchheimer Model

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**Abstract.** This study presents a comprehensive two-phase thermal model for simulating matrix acidization in porous media using the non-isothermal Darcy–Brinkman–Forchheimer framework. The model integrates multiphase flow, reactive transport, dynamic porosity evolution, and heat transfer, with temperature-dependent reaction kinetics incorporated through an Arrhenius-type formulation. A series of numerical experiments are conducted to investigate the effects of initial matrix temperature, injected acid temperature, and injection velocity on dissolution ~~behaviour~~ and wormhole formation. Results show that the initial matrix temperature has minimal influence due to rapid thermal equilibrium, while high acid temperature significantly enhances reaction rates and promote localized wormhole growth. Verification experiments confirm that increasing acid temperature produces effects similar to decreasing injection velocity, as both shift the dissolution pattern from uniform to ramified and wormhole-dominated regimes. These findings offer valuable insights for optimizing acidizing treatments by balancing thermal and hydrodynamic conditions to improve stimulation efficiency.

Deleted: behavior

## 1. Introduction

Matrix acidizing is a widely used stimulation technique in the petroleum industry aimed at enhancing the productivity of hydrocarbon reservoirs, particularly in carbonate and sandstone formations. The process involves injecting reactive acid solutions—commonly hydrochloric acid—into the reservoir at pressures below the fracture pressure, allowing the acid to flow through the porous matrix and dissolve mineral components that block pore spaces. This dissolution improves the permeability of the formation, facilitating more efficient oil and gas flow to the wellbore. Matrix acidizing is especially important in carbonate reservoirs, where it can lead to the formation of high-conductivity channels known as wormholes, significantly boosting well performance. As a cost-effective and controllable stimulation method, matrix acidizing plays a crucial role in maximizing recovery from existing wells

35 and extending the productive life of mature fields (Araújo et al., 2024; Mahdavi Kalatehno et al., 2025;  
36 Qureshi et al., 2023).

37 Numerical simulation is a crucial method for studying matrix acidizing. The evolution of numerical  
38 models for matrix acidizing can be summarized as follows: The earliest single-phase model, based on  
39 the Darcy scale, was introduced by (Daccord et al., 1993a, 1993b; Fredd and Fogler, 1998). (Golfier et  
40 al., 2002) later evaluated the Darcy-scale model's ability to capture different dissolution regimes and  
41 assess how flow parameters influence wormhole development. (Panga et al., 2005) advanced the field  
42 by proposing the two-scale model (Darcy scale and pore scale), which describes transport and reaction  
43 mechanisms in reactive dissolution and studies wormhole formation during carbonate core acidization.  
44 However, these models rely solely on Darcy's equation, which fails to accurately represent fluid  
45 dynamics in high-porosity regions that form during acidizing. As porosity increases—sometimes  
46 approaching unity in certain zones—Darcy's law becomes inadequate. Additionally, fluid velocities in  
47 these regions can rise significantly. To address this limitation, (Wu et al., 2015) introduced the Darcy-  
48 Brinkman-Forchheimer (DBF) model, incorporating Brinkman and Forchheimer terms into the  
49 momentum equations to better describe fluid dynamics in high-porosity conditions. Later, (Wu et al.,  
50 2022) combined the DBF framework with a continuum fracture model to simulate matrix acidization in  
51 fractured porous media, achieving accurate results with a simplified approach. The DBF framework has  
52 also been widely applied in other research areas, as seen in studies by (Alokaily, 2022; Deb et al., 2024;  
53 Shahid et al., 2025; Yoon and Mallikarjunaiah, 2025).

54 However, the models above assume a single water phase in the reservoir, with the acid dissolved within  
55 it. While single-phase models have been widely used in matrix acidizing studies due to their simplicity  
56 and computational efficiency, they exhibit severe limitations when applied to real reservoir conditions.  
57 A key drawback is the neglect of the oil phase, which is typically present in carbonate and sandstone  
58 formations during acidizing operations. This omission leads to an inability to capture essential two-  
59 phase flow dynamics, including capillary pressure, phase interference, and wettability-driven fluid  
60 redistribution—all of which critically influence acid accessibility and reaction efficiency (Rigi et al.,  
61 2025; Mahdavi Kalatehno et al., 2025). For example, in oil-wet systems, the presence of oil can hinder  
62 acid penetration into the matrix, modify dissolution patterns, and result in an underestimation of the  
63 pore volume required for breakthrough (Elsafih and Fahes, 2021). Additionally, single-phase models  
64 disregard saturation-dependent transport properties and chemical potential gradients, both of which  
65 play a vital role in accurately predicting wormhole development and treatment performance. While  
66 these simplified models are valuable for conceptual studies, their lack of two-phase physics limits their  
67 reliability for field-scale simulations and the optimization of acid stimulation treatments.

68 To address these challenges, researchers have developed two-phase two-scale continuum models by  
69 integrating traditional porous media flow theory with acidizing frameworks (Babaei and Sedighi, 2018;  
70 Wei et al., 2017). Notably, (Sabooniha et al., 2021) proposed a two-dimensional, two-phase model  
71 combining the Cahn–Hilliard phase-field theory with Navier–Stokes equations, solved numerically via  
72 the finite element method. Further advancing the field, (Ma et al., 2022) developed a three-dimensional  
73 two-phase acidizing model for fractured carbonate rocks, utilizing a unified pipe-network method  
74 alongside a sequential implicit time scheme and adaptive time-stepping to enhance simulation accuracy.

75 A critical challenge in carbonate acidizing is the formation of highly permeable flow channels, which  
76 significantly alter flow dynamics due to the combined effects of fluid viscosity and inertia. While most  
77 existing studies model these flows using Darcy's law alone, this approach neglects key factors such as  
78 viscous shear and inertial forces. A more robust alternative is the DBF model, which incorporates  
79 Darcy's law, the Brinkman term, and the Forchheimer correction, making it suitable for high-velocity  
80 flows and heterogeneous porous media. However, despite its advantages, the DBF model has been  
81 applied only to single-phase acidizing simulations (Kou et al., 2016, 2019; Wu and Ye, 2019), leaving  
82 its potential in two-phase systems unexplored. Thus, integrating the two-phase model with the DBF  
83 model is one contribution of this work.

84 The conventional DBF framework accounts only for mass and momentum conservation, neglecting the  
85 energy conservation equation. However, temperature variations—whether due to reservoir conditions  
86 or reaction heat—play a critical role in matrix acidizing. For instance, (Zhou et al., 2022) demonstrated  
87 that elevated temperatures significantly enhance the optimal acid injection rate, while (Jia et al., 2021)  
88 highlighted the pronounced influence of reaction temperature at higher injection rates. Their findings  
89 further indicate that the optimal injection rate decreases with increasing activation energy or decreasing  
90 acid diffusion coefficient, underscoring the necessity of incorporating thermal effects into acidizing  
91 models. To address this gap, (Wu et al., 2021) extended the single-phase DBF model by integrating the  
92 energy conservation equation, forming a thermal DBF framework capable of capturing temperature-  
93 dependent dynamics. However, this model remains limited by its single-phase assumption, restricting  
94 its applicability to real-world reservoir conditions, which typically involve multiphase flow. Therefore,  
95 this work advances the field by developing a novel two-phase thermal DBF model, representing a key  
96 contribution to the understanding of heat-fluid interactions in acidizing processes.

97 The proposed two-phase thermal DBF model addresses the limitations of traditional single-phase and  
98 isothermal frameworks by integrating several key physical mechanisms into a unified formulation. First,  
99 it incorporates multiphase momentum conservation by coupling the DBF equations with relative  
100 permeability functions, allowing for accurate representation of viscous and inertial effects in both water  
101 and oil phases. Second, it introduces energy conservation with interphase mass exchange, capturing  
102 heat transfer processes alongside acid-rock reactions and dissolution-induced phase changes. Third, the  
103 model features temperature-dependent reaction kinetics governed by Arrhenius-type relationships,  
104 enabling spatially resolved predictions of reaction rates under varying thermal conditions. Together,  
105 these components empower the model to simulate complex field phenomena with high fidelity,  
106 including thermally driven wormhole branching, rate optimization guided by thermal feedback, and  
107 quantification of competing effects between reaction enthalpy and viscous dissipation in multiphase  
108 systems. This work thus establishes the first comprehensive two-phase thermal DBF framework,  
109 effectively bridging the gap between idealized models and practical field applications where  
110 multiphase flow and thermal effects are inherently coupled. This work is organized as following.

111 Introduction: Presents the need for a thermal two-phase acidizing model to address limitations in  
112 existing single-phase or isothermal approaches.

113 Numerical Model: Formulates the coupled mass, momentum, energy, and species transport equations  
114 with temperature-dependent reaction kinetics.

115 Solution Procedure: Describes the discretization and solution procedure to implement the model.  
 116 Numerical Experiments: Shows some important conclusions from the numerical experiments.  
 117 Conclusion and Future Work: Summarizes key findings on how thermal and flow conditions govern  
 118 acidizing efficiency. Outlines extensions to heterogeneous media, larger scales, multi-mineral systems,  
 119 and optimization tools.

## 120 2. Numerical Model

121 The two-phase thermal DBF model includes the mass conservation equation for each fluid phase  $\alpha \in$   
 122  $\{w, n\}$ , which is expressed as below.  $w$  represents the wetting phase such as the water phase, and  $n$   
 123 stands for the non-wetting phase such as the oil phase.

$$124 \quad \frac{\partial(\phi S_\alpha)}{\partial t} + \nabla \cdot (\phi S_\alpha \mathbf{u}_\alpha) = 0, \quad (1)$$

125 In the equation,  $\phi$  is the porosity,  $S_\alpha$  is the saturation of phase  $\alpha$ ,  $\mathbf{u}_\alpha$  is the effective velocity of phase  $\alpha$ ,  
 126 and  $t$  is the time. This equation does not consider any source or sink term, and the fluid is  
 127 incompressible. Based on the single-phase DBF equation, the DBF equation for the two-phase  
 128 condition (the momentum conservation equation) is expressed as:

$$129 \quad \rho_\alpha \left( \frac{\partial(\phi S_\alpha \mathbf{u}_\alpha)}{\partial t} + \nabla \cdot (\phi S_\alpha \mathbf{u}_\alpha \otimes \mathbf{u}_\alpha) \right) \\ 130 \quad = -\nabla(p + \mu_\alpha) - \frac{\eta_\alpha \phi S_\alpha}{k_{r\alpha} \mathbf{K}} \mathbf{u}_\alpha + \nabla \cdot \phi S_\alpha \eta_\alpha \nabla \mathbf{u}_\alpha - \phi S_\alpha \rho_\alpha F_\alpha |\mathbf{u}_\alpha| \mathbf{u}_\alpha + \rho_\alpha \mathbf{g}, \quad (2)$$

131 where  $p$  is the fluid pressure,  $\mu_\alpha$  is the capillary pressure potential of phase  $\alpha$ ,  $\eta_\alpha$  is the dynamic  
 132 viscosity,  $\mathbf{K}$  is the absolute permeability,  $k_{r\alpha}$  is the relative permeability,  $\rho_\alpha$  is the mass density of  
 133 phase  $\alpha$ ,  $F_\alpha$  is the Forchheimer drag coefficient, and  $\mathbf{g}$  is the gravity. The fluid is assumed as the  
 134 Newtonian fluids (constant viscosity). The chemical potential  $\mu_\alpha$  due to capillary effects is defined as  
 135 (Kou et al., 2021,2023,2024):

$$136 \quad \mu_w = \gamma_w \ln(S_w) + \gamma_{wn} S_n, \quad (3)$$

$$137 \quad \mu_n = \gamma_n \ln(S_n) + \gamma_{wn} S_w, \quad (4)$$

138 with the saturation constraint

$$139 \quad S_w + S_n = 1, \quad (5)$$

140 where  $\gamma_w$ ,  $\gamma_n$ , and  $\gamma_{wn}$  are phenomenological constants related to fluid-fluid interactions. The transport  
 141 of reactant (acid) concentration  $C_f$  in the wetting phase is governed by:

$$142 \quad \frac{\partial(\phi S_w C_f)}{\partial t} + \nabla \cdot (\phi S_w \mathbf{u}_w C_f) - \nabla \cdot (\phi S_w \mathbf{D}_e \cdot \nabla C_f) = -\frac{S_w a_v k_s k_c C_f}{k_c + k_s}, \quad (6)$$

143 where  $\mathbf{D}_e$  is the effective diffusivity,  $a_v$  is the specific surface area per unit volume,  $k_c$ ,  $k_s$  are reaction  
 144 rate constants for mass transfer and surface reaction, respectively. The porosity-progress equation is  
 145 modeled as:

$$146 \quad \frac{\partial \phi}{\partial t} = \frac{\alpha S_w a_v k_s k_c C_f}{\rho_s (k_c + k_s)}, \quad (7)$$

147 where  $\rho_s$  is the solid matrix density,  $\alpha$  is a stoichiometric coefficient.

148 All the equations above are established with the constant-temperature assumption. However, it is  
 149 known that temperature is an important factor to influence the chemical reaction in reservoirs. Thus, to  
 150 consider the temperature effect, we introduce the energy conservation equation as following.

$$\begin{aligned}
 151 \quad & \sum_{\alpha=w,n} \rho_{\alpha} \theta_{\alpha} \frac{\partial(\phi S_{\alpha} T)}{\partial t} + \rho_s \theta_s \frac{\partial((1-\phi)T)}{\partial t} + \sum_{\alpha=w,n} \rho_{\alpha} \theta_{\alpha} \nabla \cdot (\phi S_{\alpha} \mathbf{v}_{\alpha} T) \\
 152 \quad & - \nabla \cdot (\phi S_w M_w + \phi S_n M_n + (1-\phi) M_s) \nabla T \\
 153 \quad & = -2H_r a_v k_e C_f + \sum_{\alpha=w,n} \left( \frac{\eta_{\alpha} \phi S_{\alpha}}{k_r \alpha K} |\mathbf{v}_{\alpha}|^2 + \eta_{\alpha} \phi S_{\alpha} |\nabla \mathbf{v}_{\alpha}|^2 + \phi S_{\alpha} \rho_{\alpha} F_{\alpha} |\mathbf{v}_{\alpha}|^3 \right), \quad (8)
 \end{aligned}$$

154 where  $T$  stands for the temperature,  $\theta_{\alpha}$  is the specific heat capacity of the phase  $\alpha$ ,  $M_{\alpha}$  is the thermal  
 155 conductivity of the phase  $\alpha$ .  $H_r$  is the reaction heat, which is expressed as

$$156 \quad H_r = -6846 + 8.038T - 0.00322T^2 - 870.3T^{-1}. \quad (9)$$

157 Besides that,  $k_e$  is a coefficient, which is defined as

$$158 \quad k_e = \frac{2k_c k_s}{k_c + 2k_s}. \quad (10)$$

159 The value of the molecular diffusion coefficient  $d_m$  depends on the temperature directly, which is  
 160 called the Arrhenius-type dependencies. It can be expressed as

$$161 \quad d_m = d_{m0} \cdot e^{\frac{E_g}{R_g} \left( \frac{1}{T_0} - \frac{1}{T} \right)}. \quad (11)$$

162 The subscript “0” stands for the initial value.  $E_g$  is the activation energy, and  $R_g$  is the molar gas  
 163 constant. Moreover,  $k_s$  also has the Arrhenius-type dependencies, which is expressed as

$$164 \quad k_s = k_{s0} \cdot e^{\frac{E_g}{R_g} \left( \frac{1}{T_0} - \frac{1}{T} \right)}. \quad (12)$$

165 By the temperature-dependent parameters above, the reaction term in the concentration equation (6)  
 166 and the porosity equation (7) are affected by the temperature.

167 The equations form a fully coupled, nonlinear system that integrates two-phase flow dynamics,  
 168 momentum transport, reactive acid dissolution, porosity evolution, and temperature changes to  
 169 accurately simulate wormhole formation during acidizing operations. Temperature effects are critically  
 170 incorporated through Arrhenius-type dependencies for molecular diffusion ( $d_m$ ) and surface reaction  
 171 rates ( $k_s$ ), which directly influence acid transport (via the acid concentration equation) and porosity  
 172 development (via the porosity-progress equation). The evolving porosity then feeds back into the mass  
 173 and momentum conservation equations by modifying permeability and pore structure parameters, while  
 174 the computed flow velocities and reaction rates determine heat generation in the energy conservation  
 175 equation. This bidirectional coupling creates a continuous feedback loop: updated temperature values  
 176 alter reaction kinetics and transport properties, which in turn affect fluid flow and dissolution patterns.  
 177 The model's ability to capture these complex thermo-hydro-chemical interactions enables precise  
 178 prediction of wormhole propagation under realistic reservoir conditions, providing valuable insights for  
 179 optimizing stimulation treatments through comprehensive parameter sensitivity analysis and  
 180 performance evaluation.

181 In the system,  $S_w$ ,  $\phi$ ,  $p$ ,  $\mathbf{u}_\alpha$ ,  $C_f$  and  $T$  are the main unknowns to be solved by the numerical schemes  
 182 mentioned later. The other variables can be computed by the following equations. The Forchheimer  
 183 drag coefficient is calculated as

$$184 \quad F_\alpha = \frac{1.75}{\sqrt{150(S_\alpha \phi)^3}}. \quad (13)$$

185 The relative permeability is

$$186 \quad k_{r\alpha} = S_\alpha^3. \quad (14)$$

187  $\mathbf{D}_e$  is expressed as

$$188 \quad \mathbf{D}_e = \alpha_{OS} d_m \mathbf{I} + \frac{\|\mathbf{u}_w\|}{\phi S_w} (d_l \mathbf{E} + d_t \mathbf{E}^\perp). \quad (15)$$

189 In 2D condition,

$$190 \quad \mathbf{E} = \frac{1}{\|\mathbf{u}_w\|^2} \begin{pmatrix} u_{wx}^2 & u_{wx}u_{wy} \\ u_{wy}u_{wx} & u_{wy}^2 \end{pmatrix}, \quad (16)$$

$$191 \quad \mathbf{E}^\perp = \mathbf{I} - \mathbf{E}. \quad (17)$$

192  $u_{wx}$ ,  $u_{wy}$  stand for the wetting-phase velocity of the  $x$ - and  $y$ -directions, respectively.  $d_l$  and  $d_t$  stand  
 193 for the longitudinal dispersion coefficient and the transverse dispersion coefficient, respectively.

$$194 \quad d_l = \alpha_{OS} d_m + \frac{2\lambda_x \|\mathbf{u}\| r_p}{\phi}, \quad (18)$$

$$195 \quad d_t = \alpha_{OS} d_m + \frac{2\lambda_T \|\mathbf{u}\| r_p}{\phi}. \quad (19)$$

196  $\alpha_{OS}$  is a constant depending on pore connectivity.  $\lambda_x$  and  $\lambda_T$  are constants depending on the structure  
 197 of the medium. Besides that, there are Kozeny-Carman equations as below

$$198 \quad \frac{K}{K_0} = \frac{\phi}{\phi_0} \left( \frac{\phi(1-\phi_0)}{\phi_0(1-\phi)} \right)^2, \quad (20)$$

$$199 \quad \frac{r_p}{r_0} = \sqrt{\frac{K\phi_0}{K_0\phi}}, \quad (21)$$

$$200 \quad \frac{\alpha_v}{\alpha_0} = \frac{\phi r_0}{\phi_0 r_p}. \quad (22)$$

201 If the matrix is homogeneous, the permeability value  $K$  can be represented as a scalar. Based on Eq.  
 202 (20),  $K$  can be directly calculated once  $\phi$  is known. Furthermore, by combining Eqs. (20) - (22), the  
 203 parameter  $\alpha_v$  can be expressed as a function of  $\phi$ , enabling its determination.  $r_p$  is the pore radius.  $k_c$  is  
 204 calculated from Eq. (23).

$$205 \quad k_c = \frac{d_m}{2r_p} \left( Sh_\infty + 0.7 \left( \frac{2\|\mathbf{u}_w\| r_p \rho_w}{\eta_w} \right)^{\frac{1}{2}} \left( \frac{\eta_w d_m}{\rho_w} \right)^{\frac{1}{3}} \right), \quad (23)$$

206 where  $Sh_\infty$  is the asymptotic Sherwood number.

### 207 3. Solution Procedure

208 The strongly coupled nonlinear equation system presents significant computational challenges due to  
 209 its inherent complexity. To address this, we employ a structured numerical approach using a uniform  
 210 grid with spatial resolutions  $\Delta x$  and  $\Delta y$  in the respective coordinate directions, along with uniform time

211 steps  $\Delta t$  for temporal discretization. The solution adopts a staggered grid arrangement, with primary  
 212 variables (water saturation  $S_w$ , porosity  $\phi$ , pressure  $p$ , acid concentration  $C_f$ , and temperature  $T$ ) stored  
 213 at cell centers while phase velocities  $\mathbf{u}_\alpha$  are defined at cell edges. Spatial approximations are carefully  
 214 implemented: harmonic averaging is applied for edge porosity~~s~~ and permeability~~s~~ using adjacent cell-  
 215 center values, while arithmetic averaging is used for nodal porosity and saturation calculations. The  
 216 scheme employs upwind discretization for both edge saturations and acid concentrations to ensure  
 217 numerical stability. The gradients can be derived with the finite difference method. This conservative  
 218 formulation maintains proper mass balance while effectively handling the complex coupling between  
 219 multiphase flow, reactive transport, evolving porosity-permeability relationships, and temperature  
 220 change that characterize the acidizing process.

221 To solve the coupled system, it is first necessary to define a consistent solution sequence for the  
 222 primary unknowns. The porosity  $\phi^{n+1}$  is updated first, as it is directly governed by the porosity  
 223 evolution equation (Eq. (7)), which is relatively straightforward. A first-order backward Euler scheme  
 224 is used for time discretization throughout the model. Applying this scheme to Eq. (7) allows for the  
 225 explicit computation of the updated porosity at the beginning of each time step, where the superscript  $n$   
 226 denotes the previous time level. Once porosity is updated, the next step is to solve for the wetting-phase  
 227 saturation  $S_w^{n+1}$ . This requires discretizing the wetting-phase mass conservation equation (Eq. (1)),  
 228 which leads to a linear system in which  $S_w^{n+1}$  is the unknown. Solving this system yields the new  
 229 saturation field. Following this, the model proceeds to solve for pressure  $p^{n+1}$  and phase velocities  
 230  $\mathbf{u}_\alpha^{n+1}$  simultaneously, using a coupled formulation of the momentum and mass conservation equations.  
 231 To do so, the mass conservation equations for both wetting and non-wetting phases are summed to  
 232 yield a unified continuity equation:

$$233 \quad \frac{\partial \phi}{\partial t} + \sum_{\alpha=w,n} S_\alpha (\nabla \cdot (\phi S_\alpha \mathbf{u}_\alpha)) = 0. \quad (24)$$

234 This equation, in combination with the momentum equation (Eq. (2)), is discretized to form a linear  
 235 system in which the unknowns are  $p^{n+1}$  and  $\mathbf{u}_\alpha^{n+1}$ . Since the DBF momentum equation is nonlinear in  
 236 velocity, appropriate linearization is performed: the advective term is approximated as  $\nabla \cdot$   
 237  $(\phi S_\alpha \mathbf{u}_\alpha^n \otimes \mathbf{u}_\alpha^{n+1})$ , while the Forchheimer term is linearized as  $\phi S_\alpha \rho_\alpha F_\alpha |\mathbf{u}_\alpha^n| \mathbf{u}_\alpha^{n+1}$ . Once the velocity  
 238 and pressure fields are updated, the acid concentration  $C_f^{n+1}$  is solved by discretizing the reactive  
 239 transport equation (Eq. (6)), again yielding a linear system. The reaction source term is treated  
 240 implicitly using values from the new time step. The final step in the solution sequence is the update of  
 241 temperature  $T^{n+1}$ , governed by the energy conservation equation (Eq. (8)). Here, all temperature-  
 242 dependent terms on the left-hand side are discretized implicitly (using  $T^{n+1}$ ), while the source and  
 243 transport terms on the right-hand side are discretized explicitly (using  $T^n$ ). This approach yields a  
 244 linear system for  $T^{n+1}$ , which can be solved to update the thermal field. This time-stepping procedure  
 245 is repeated iteratively until either the final simulation time is reached or breakthrough occurs.  
 246 Breakthrough is defined as the moment when the pressure drop between the inlet and outlet boundaries  
 247 decreases to within 1% of its initial value, indicating the formation of a high-conductivity channel. It is

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251 important to ensure that all auxiliary variables are updated prior to their use in the computation of the  
 252 primary unknowns.  
 253 High-resolution simulation of matrix acidizing in porous media requires fine computational grids to  
 254 accurately resolve porosity heterogeneity and capture wormhole dynamics. However, the resulting  
 255 large-scale systems significantly increase computational demand. To address this, an efficient parallel  
 256 computing strategy is developed, focusing on domain decomposition, inter-processor communication,  
 257 and distributed solution of the coupled linear system. More details on the parallelization can be found  
 258 in (Wu et al., 2015).

#### 259 4. Numerical Experiments

##### 260 4.1 Experiments with different initial matrix temperatures

261 To investigate the influence of the initial matrix temperature on two-phase acidizing, three numerical  
 262 experiments on single-reactive-mineral carbonate matrix are conducted, each with a constant acid  
 263 injection temperature but different initial matrix temperatures. The injection temperature of the acid is  
 264 maintained at 320 K, while the initial matrix temperatures are set to 320 K, 420 K, and 520 K,  
 265 respectively. The computational domain is a rectangular matrix measuring 0.1 m × 0.1 m, discretized  
 266 into a uniform grid of 120 × 120 cells. Initially, the matrix is nearly saturated with oil, with a water  
 267 saturation of 0.01. Acid is injected from the left boundary at a constant velocity of  $4.17 \times 10^{-5}$  m/s,  
 268 resulting in a water saturation of 1.0 at that boundary. The injected acid has a concentration of 500  
 269 mol/m<sup>3</sup>. The right boundary serves as the outflow, while the top and bottom boundaries are  
 270 impermeable. To represent natural heterogeneity, the initial porosity is spatially randomized with an  
 271 average value of 0.18. To that end, the function “random number (rand)” in FORTRAN90 is used to  
 272 generate an array of pseudorandom numbers from the uniform distribution over the range [0, 1]. Then,  
 273 the numbers in [0, 1] is mapped to another range [0.15, 0.21] by the formulation: new rand =  
 274 0.06\*rand + 0.15. The average value of new rand in [0.15, 0.21] is 0.18. Gravity is applied in the y-  
 275 direction. The relevant physical and chemical parameters used in the simulations are summarized in  
 276 Table 1. The simulation is performed with a time step of ten seconds, and the UMFPAK solver is  
 277 employed for solving the resulting linear systems.

278 **Table 1 Physical and chemical parameters of the experiments**

Parameter	Value	Unit
$\alpha_{v0}$	$5.0 \times 10^{-1}$	m <sup>-1</sup>
$C_{f,in}$	$5.0 \times 10^2$	mol/m <sup>3</sup>
$d_{m0}$	$3.6 \times 10^{-9}$	m <sup>2</sup> /s
$g_y$	-9.807	m/s <sup>2</sup>
$K_0$	$9.869233 \times 10^{-16}$	m <sup>2</sup>
$k_{s0}$	$2.0 \times 10^{-3}$	m/s
$M_n$	$1.5 \times 10^1$	W/(m · K)

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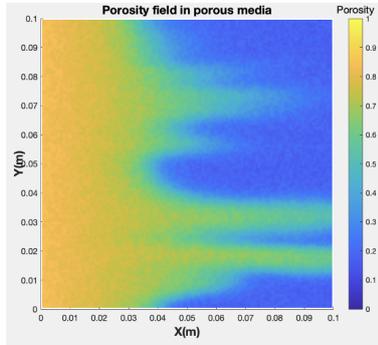
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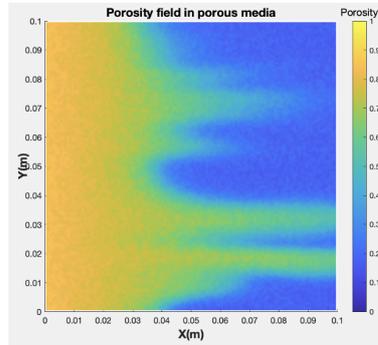
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$M_s$	$5.526 \times 10^2$	W/(m · K)
$M_w$	$6.0 \times 10^1$	W/(m · K)
$r_0$	$1.0 \times 10^{-6}$	m
$Sh_\infty$	3.66	
$\alpha$	$5.0 \times 10^{-2}$	kg/mol
$\alpha_{OS}$	$5.0 \times 10^{-1}$	
$\gamma_n$	0.0	Pa · s
$\gamma_w$	$5.8 \times 10^2$	Pa · s
$\gamma_{wn}$	0.0	Pa · s
$\eta_n$	$1.0 \times 10^{-2}$	Pa · s
$\eta_w$	$1.0 \times 10^{-3}$	Pa · s
$\theta_n$	$2.0 \times 10^2$	J/(kg · K)
$\theta_s$	$2.0 \times 10^1$	J/(kg · K)
$\theta_w$	$4.184 \times 10^2$	J/(kg · K)
$\lambda_T$	$1.0 \times 10^{-1}$	
$\lambda_X$	$5.0 \times 10^{-1}$	
$\rho_n$	$9.0 \times 10^2$	kg/m <sup>3</sup>
$\rho_s$	$2.71 \times 10^3$	kg/m <sup>3</sup>
$\rho_w$	$1.01 \times 10^3$	kg/m <sup>3</sup>
$\Phi_0$	$1.8 \times 10^{-1}$	

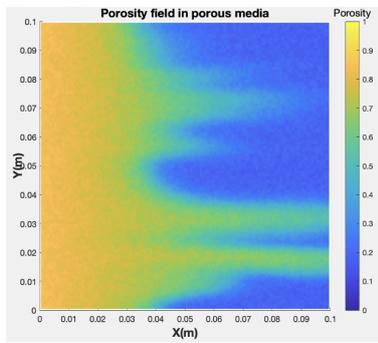
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(a) Porosity profile with initial matrix temperature of 320 K.



(b) Porosity profile with initial matrix temperature of 420 K.



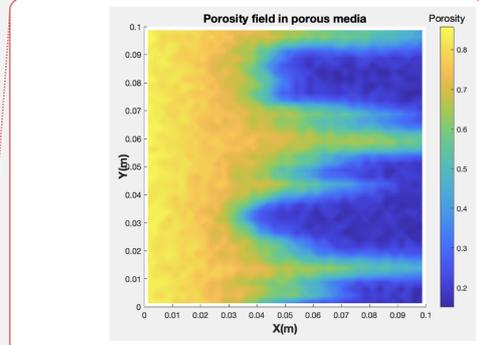
(c) Porosity profile with initial matrix temperature of 520 K.

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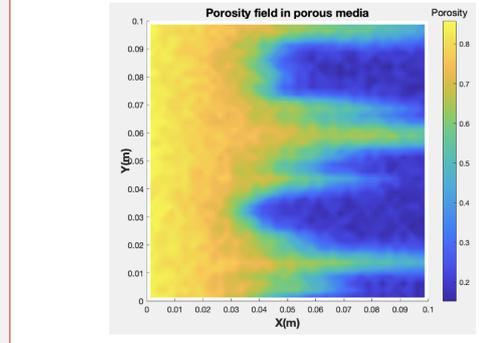
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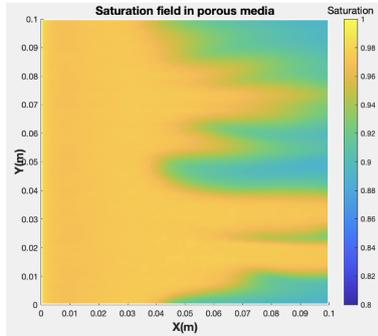
**Figure 1 Porosity profiles of the two-phase experiments at breakthrough with different initial matrix temperatures.**



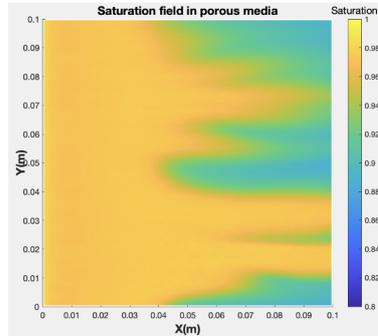
(a) Porosity profile with initial matrix temperature of 320 K.



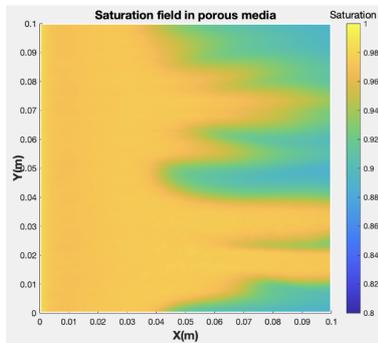
Deleted: (c) Porosity profile with initial matrix temperature of 520 K.



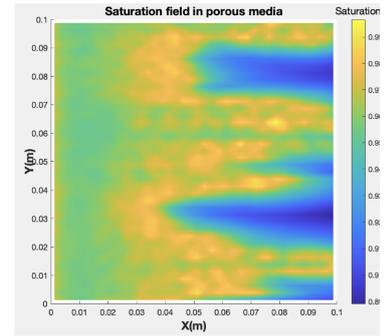
(a) Saturation profile with initial matrix temperature of 320 K.



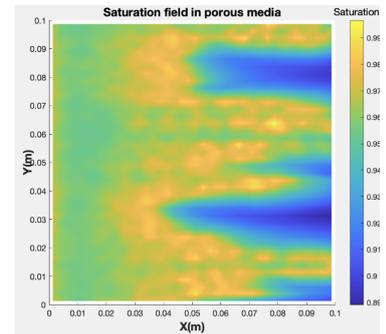
(b) Saturation profile with initial matrix temperature of 420 K.



(c) Saturation profile with initial matrix temperature of 520 K.

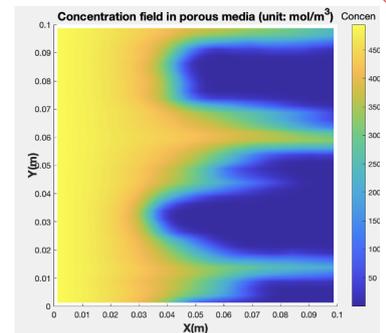


(a) Saturation profile with initial matrix temperature of 320 K.

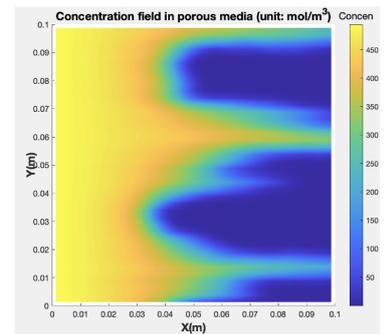


(c) Saturation profile with initial matrix temperature of 520 K.

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(a) Concentration profile with initial matrix temperature of 320 K.



(c) Concentration profile with initial matrix temperature of 520 K.

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**Figure 2 Saturation profiles of the two-phase experiments at breakthrough with different initial matrix temperatures.**

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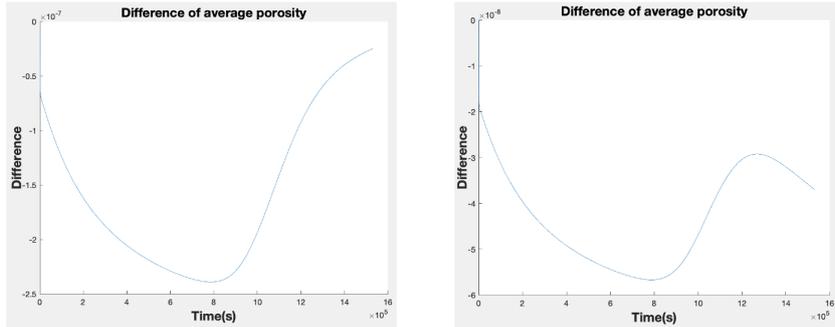
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The numerical simulations reveal that variations in the initial matrix temperature—set at 320 K, 420 K, and 520 K—do not lead to significant differences in the outcome of the two-phase acidizing process. Across all three cases, the acid is injected at a constant temperature of 320 K, and the simulation is run until breakthrough, which occurs when the pressure differential between inlet and outlet stabilizes within 1% of its initial value. The computed pore volume to breakthrough (PVBT) is 6.392, in all scenarios, suggesting that the overall rate of acid propagation and dissolution behavior is largely insensitive to the initial thermal state of the matrix.

This conclusion can also be drawn from Figure 1 and Figure 2. The two figures show that the porosity and saturation profiles at breakthrough are highly similar across the three temperature settings. The acid-induced dissolution patterns develop with nearly identical wormhole morphologies, and the acid front progresses in a comparable manner through the porous medium. This indicates that, under the given conditions, the chemical reaction kinetics and transport processes are predominantly governed by the injected acid properties and flow dynamics, rather than the initial temperature of the solid matrix.

However, the little differences of the simulation results among different initial matrix temperature in Figure 1 and Figure 2 cannot be clarified by eyeballs. Thus, the differences in average porosity between different initial matrix temperatures during the simulation are given in Figure 3. From the

338 figure, it is clearly seen that the differences between different initial matrix temperatures are in the  
 339 order of  $10^{-7}$  and  $10^{-8}$ , respectively, which demonstrates the negligible influence of initial matrix  
 340 temperatures on acidizing.  
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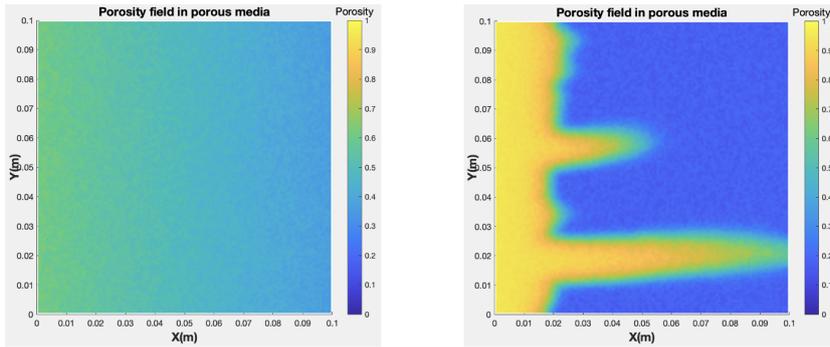


(a) The difference in average porosity between matrices with initial temperatures of 320 K and 420 K. (b) The difference in average porosity between matrices with initial temperatures of 420 K and 520 K.

342 **Figure 3 Differences in average porosity between different initial matrix temperatures during the**  
 343 **simulation.**  
 344

#### 345 4.2 Experiments with different acid temperatures

346 Since the acidizing process shows minimal sensitivity to the initial matrix temperature, this section  
 347 focuses on investigating the influence of the injected acid temperature on the acidizing behavior. The  
 348 initial matrix temperature is fixed at 420 K, while the injection temperatures are varied, set at 290 K  
 349 and 360 K, respectively. It is noted that the range of the acid temperature is generally between 273 K  
 350 and 373 K in real applications, and therefore the acid temperatures outside the range are not  
 351 considered. The simulation is performed with a time step of one second. All other experimental  
 352 conditions remain largely consistent with those described in the previous section.



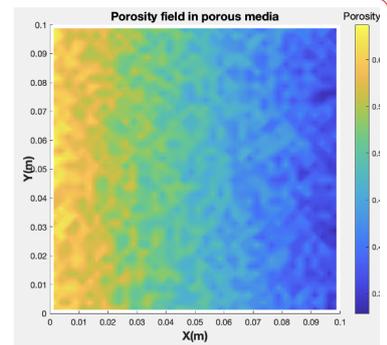
(a) Porosity profile with initial matrix temperature of 290 K. (b) Porosity profile with initial matrix temperature of 360 K.

353 **Figure 4 Porosity profiles of the two-phase experiments at breakthrough with different acid**  
 354 **temperatures.**  
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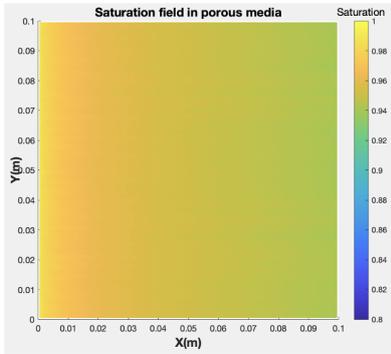
~~Deleted:~~ Similarly, Figure 4 and Figure 5, which present the streamline patterns of the water and oil phases, show consistent flow structures in all cases. The injected acid solution follows the same preferential paths, displacing oil and generating high-permeability channels in a nearly identical fashion. The similarity of these flow patterns reinforces the observation that the influence of initial matrix temperature is minimal in this two-phase system, at least under the conditions studied.

Finally, the temperature profiles shown in Figure 6 confirm that although the initial matrix temperature is different, the overall thermal field at breakthrough converges toward a similar distribution. This is likely due to the dominance of the injected fluid temperature (320 K) and its thermal diffusion into the matrix. As a result, the initial temperature differences are progressively diminished during the acidizing process. Overall, the results suggest that for this set of parameters, the initial matrix temperature has a negligible effect on the outcome of the acidizing process, including wormhole formation, fluid displacement, and breakthrough characteristics. This points to a degree of thermal robustness in the system, which could simplify operational strategies in real applications where subsurface temperature variations are present.

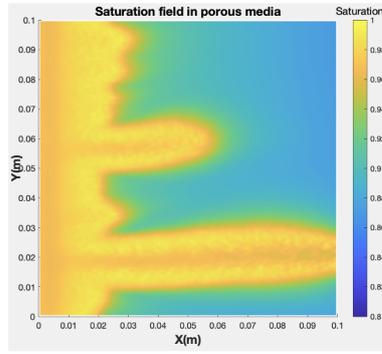


~~Deleted:~~ (a) Porosity profile with initial matrix temperature of 290 K.

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(a) Saturation profile with initial matrix temperature of 290 K.



(b) Saturation profile with initial matrix temperature of 360 K.

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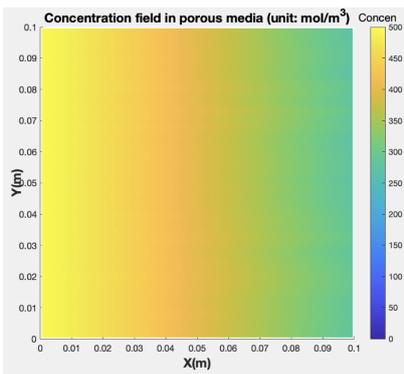
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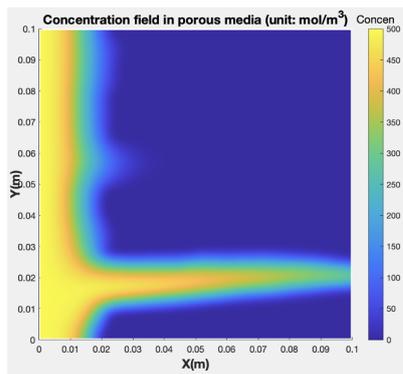
**Figure 5** Saturation profiles of the two-phase experiments at breakthrough with different acid temperatures.

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(a) Concentration profile with initial matrix temperature of 290 K.



(b) Concentration profile with initial matrix temperature of 360 K.

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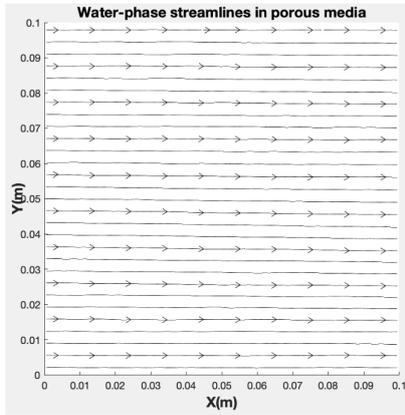
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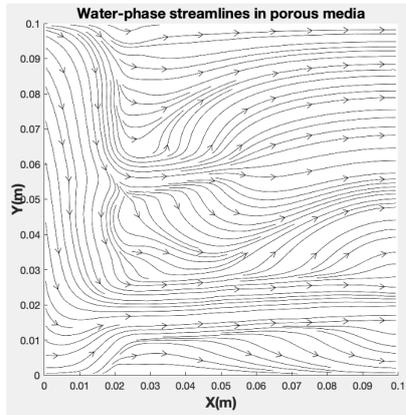
**Figure 6** Concentration profiles of the two-phase experiments at breakthrough with different acid temperatures.

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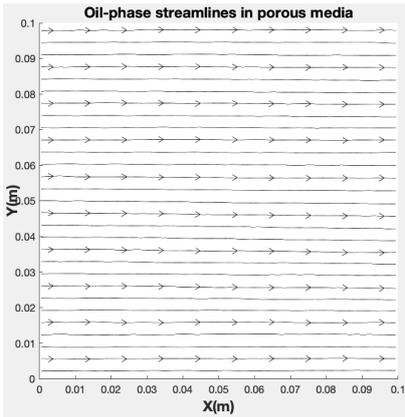


(a) Water-phase streamlines with initial matrix temperature of 290 K.

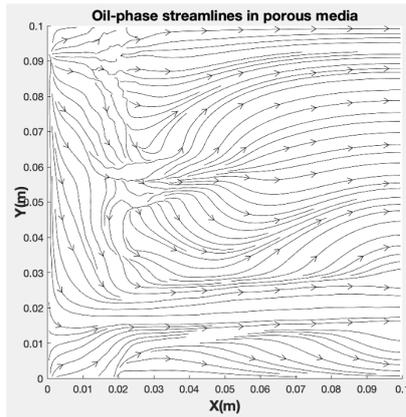


(b) Water-phase streamlines with initial matrix temperature of 360 K.

**Figure 7** Water-phase streamlines of the two-phase experiments at breakthrough with different acid temperatures.

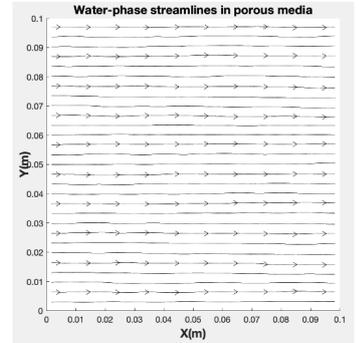


(a) Oil-phase streamlines with initial matrix temperature of 290 K.



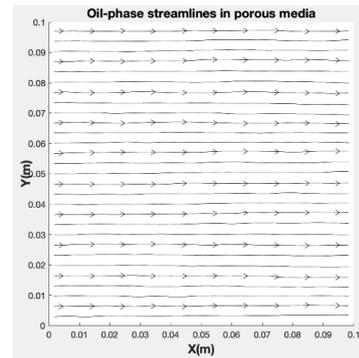
(b) Oil-phase streamlines with initial matrix temperature of 360 K.

**Figure 8** Oil-phase streamlines of the two-phase experiments at breakthrough with different acid temperatures.



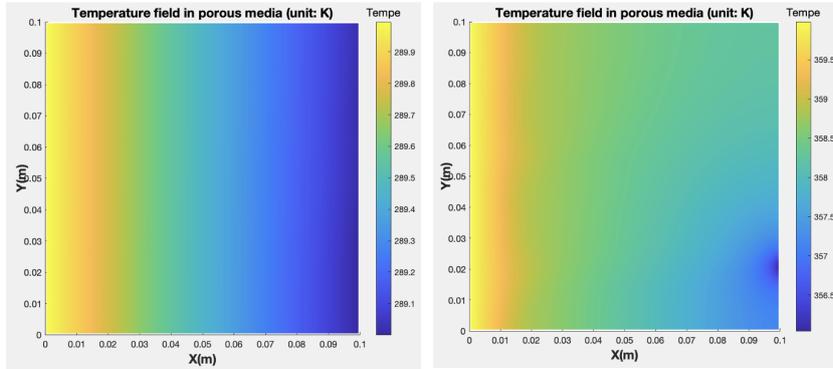
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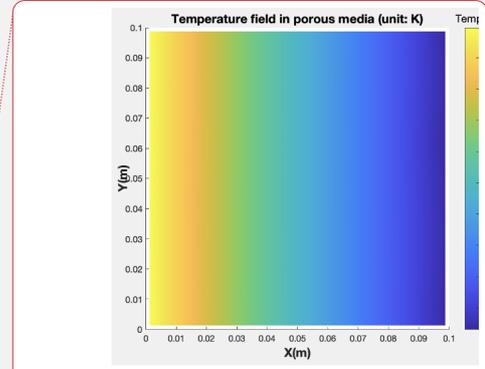


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(a) Temperature profile with initial matrix temperature of 290 K. (b) Temperature profile with initial matrix temperature of 360 K.



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**Figure 9. Temperature profiles of the two-phase experiments at breakthrough with different acid temperatures.**

This section explores the effect of injected acid temperature on the two-phase acidizing process while keeping the initial matrix temperature constant at 420 K. Three scenarios are examined: acid injection temperatures of 290 K, 320 K (from Section 4.1), and 360 K. The pore volume to breakthrough (PVBT) values corresponding to these cases are 10.92, 6.39, and 3.12, respectively. The results indicate a strong dependence of acidizing efficiency on the injected acid temperature. As the injection temperature increases, the PVBT significantly decreases, implying faster breakthrough and more efficient acid propagation through the matrix. This trend is attributed to the temperature dependence of the reaction kinetics—higher injection temperatures accelerate the acid–rock reaction rate, leading to faster porosity development, more rapid wormhole formation, and consequently earlier breakthrough.

Figure 4 and Figure 5 present the porosity and saturation profiles at breakthrough for acid injection temperatures of 290 K and 360 K. At 360 K, dissolution is highly localized and concentrated, forming a well-developed wormhole structure. In contrast, the 290 K case exhibits a more diffuse and widespread dissolution front, indicative of slower reaction kinetics and reduced channeling efficiency. This behavior corresponds to a uniform dissolution pattern, where acid interacts more evenly with the porous matrix rather than focusing along a dominant path. The corresponding saturation profiles further support this distinction. At 360 K, water displacement is strongly aligned with the main flow path, producing a narrow and well-defined saturation front. Conversely, the 290 K case shows a broader and more dispersed water front, reflecting the less focused nature of fluid transport under cooler acid injection conditions. For the intermediate case of 320 K (as discussed in Section 4.1), the porosity profile lies between the 290 K and 360 K cases. It displays several small branching fingers extending from the injection boundary, characteristic of a ramified dissolution pattern. According to (Wu, 2015), acidizing can produce five primary dissolution patterns—face, conical, wormhole, ramified, and uniform—depending primarily on the increase of the injection velocity, assuming other conditions remain constant. Interestingly, the observed trend suggests that increasing the injected acid temperature produces similar effects to decreasing the injection velocity, both leading to a transition from uniform toward wormhole-type dissolution. This highlights the dual role of thermal and hydrodynamic controls

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438 in shaping the acidizing outcome and suggests that temperature can be used as an alternative lever to  
439 optimize dissolution behavior in two-phase systems.

440 The acid concentration fields (Figure 6) further support these findings. At lower acid temperatures,  
441 acid is consumed more gradually and over a broader region, resulting in a more uniform concentration  
442 distribution. At 360 K, acid is rapidly consumed along the high-permeability wormhole, leading to a  
443 sharp depletion front and improved reactant utilization efficiency. The streamline patterns (Figure 7,  
444 and Figure 8) also reflect the temperature-dependent channelization behavior. For the higher acid  
445 temperature (360 K), streamlines are strongly aligned with the wormhole path, concentrating both  
446 water and oil flow into narrow regions. This results in more focused displacement and enhanced acid  
447 penetration. At 290 K, streamlines are more dispersed, suggesting a less efficient fluid transport and  
448 weaker coupling between dissolution and flow. Finally, temperature profiles (Figure 9) illustrate how  
449 injected acid influences the thermal distribution in the matrix. At 360 K, the elevated injection  
450 temperature maintains a thermal gradient that sustains high reaction rates along the main channel. In  
451 contrast, the 290 K injection case shows a cooler and more diffused thermal field, contributing to lower  
452 reaction intensity. It is noted that the temperature scales are different for the two cases in Figure 9. If a  
453 unified temperature scale is given, the change of the temperature cannot be identified clearly.

454 In summary, the simulation results clearly demonstrate that the injected acid temperature plays a  
455 critical role in the efficiency of the acidizing process. Higher injection temperatures lead to accelerated  
456 reaction rates, more pronounced wormhole formation, and reduced PVBT values, indicating more  
457 efficient usage of the injected fluid. This finding highlights the importance of thermal management in  
458 designing effective acidizing strategies in two-phase flow systems.

### 459 4.3 Verification experiments

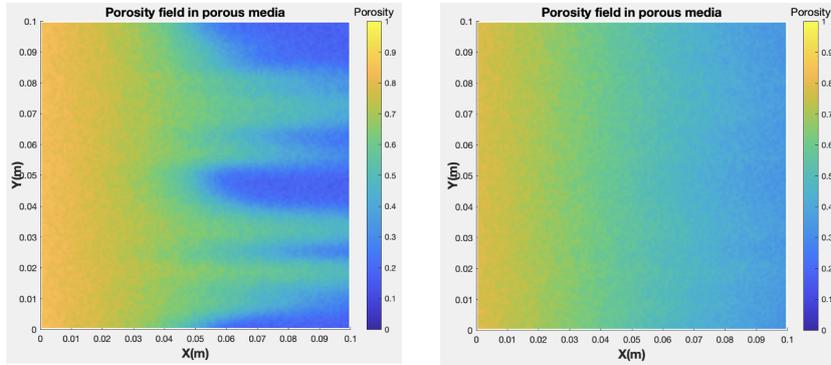
460 As discussed above, increasing the injected acid temperature appears to have an effect analogous to  
461 decreasing the injection velocity, particularly in terms of dissolution pattern and breakthrough behavior.  
462 To further examine this relationship, two additional simulations are conducted with varying injection  
463 velocities. In all previous experiments, the injection velocity was fixed at  $4.17 \times 10^{-5}$  m/s. In the new  
464 cases, the injection velocity is increased to  $4.17 \times 10^{-4}$  m/s and  $1.0 \times 10^{-3}$  m/s, respectively. The  
465 injected acid temperature is maintained at 360 K, and the initial matrix temperature remains at 420 K  
466 for all scenarios. Thus, under the fixed thermal conditions, three different injection velocities are  
467 considered:  $4.17 \times 10^{-5}$  m/s,  $4.17 \times 10^{-4}$  m/s, and  $1.0 \times 10^{-3}$  m/s. To keep numerical stability, the  
468 simulation time steps are decreased to 0.5 second and 0.1 second for the injection velocities of  $4.17 \times$   
469  $10^{-4}$  m/s and  $1.0 \times 10^{-3}$  m/s, respectively. The simulation results corresponding to the lowest velocity  
470 have already been presented in Section 4.2. The results for the two higher-velocity cases are provided  
471 and discussed in the following section.

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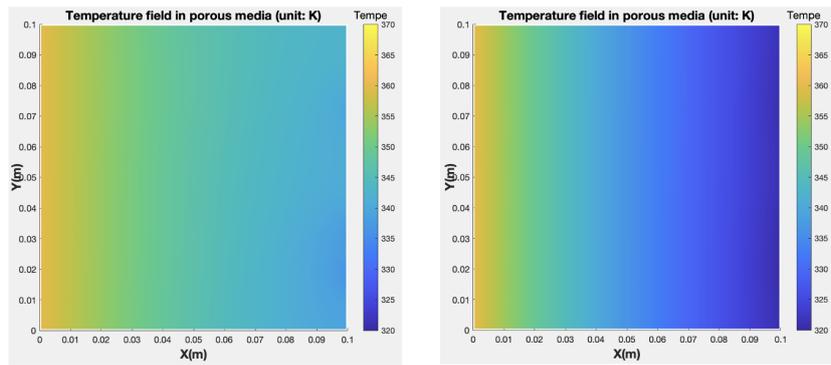
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(a) Porosity profile with injection velocity of  $4.17 \times 10^{-4}$  m/s. (b) Porosity profile with injection velocity of  $1.0 \times 10^{-3}$  m/s.

476

477 **Figure 10** Porosity profiles of the two-phase experiments at breakthrough with different injection  
478 velocity.

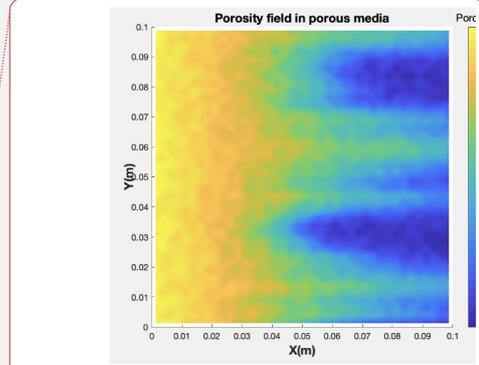


(a) Temperature profile with injection velocity of  $4.17 \times 10^{-4}$  m/s. (b) Temperature profile with injection velocity of  $1.0 \times 10^{-3}$  m/s.

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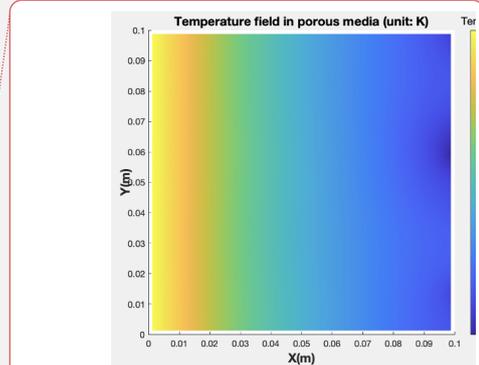
480 **Figure 11** Temperature profiles of the two-phase experiments at breakthrough with different  
481 injection velocity.

482 The porosity fields at breakthrough (Figure 10) reveal a distinct shift in the dissolution pattern as the  
483 injection velocity increases. At an injection velocity of  $4.17 \times 10^{-4}$  m/s, the dissolution exhibits a  
484 ramified pattern, characterized by multiple branching wormholes extending from the injection  
485 boundary. This morphology reflects a moderately efficient reaction front where several channels  
486 compete and develop simultaneously. In contrast, at the higher injection velocity of  $1.0 \times 10^{-3}$  m/s, the  
487 dissolution becomes more uniform, with acid spreading broadly across the domain and minimal  
488 evidence of localized channeling. This suggests that the elevated flow rate dilutes the acid-rock  
489 interaction front, preventing the formation of focused wormholes and instead promoting a diffuse  
490 dissolution regime. These observed transitions align well with the established classification of  
491 dissolution patterns (face  $\rightarrow$  conical  $\rightarrow$  wormhole  $\rightarrow$  ramified  $\rightarrow$  uniform) as a function of increasing  
492 injection velocity (Wu, 2015). Specifically, the simulation with  $4.17 \times 10^{-4}$  m/s corresponds to the  
493 ramified regime, while the simulation with  $1.0 \times 10^{-3}$  m/s corresponds to the uniform regime. These



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Deleted: (a) Temperature profile with injection velocity of  $4.17 \times 10^{-4}$  m/s.

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499 findings reinforce the earlier conclusion that increasing the injected acid temperature has an effect  
 500 analogous to decreasing the injection velocity, as both strategies favor more localized dissolution and  
 501 efficient wormhole formation. The observations are summarized in **Table 2**,  
 502 The temperature fields (**Figure 11**) show that in both cases, the matrix is thermally dominated by the  
 503 injected acid at 360 K, which progressively displaces the initially hotter matrix temperature of 420 K.  
 504 The primary distinction lies in the temperature gradient: the simulation with  $1.0 \times 10^{-3}$  m/s exhibits a  
 505 wider range between the maximum and minimum temperatures, indicating a more pronounced  
 506 temperature gradient. This suggests that higher injection velocities enhance advective heat transport,  
 507 leading to sharper thermal fronts across the domain.

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508 **Table 2 Dissolution patterns in different thermal and hydrodynamic conditions**

Dissolution pattern	$T_{\text{injection acid}} = 360 \text{ K}$	$v_{\text{injection acid}} = 4.17 \times 10^{-5} \text{ m/s}$
	$T_{\text{initial matrix}} = 420 \text{ K}$	$T_{\text{initial matrix}} = 420 \text{ K}$
<b>Uniform</b>	$v_{\text{injection acid}} = 1.0 \times 10^{-3} \text{ m/s}$	$T_{\text{injection acid}} = 290 \text{ K}$
<b>Ramified</b>	$v_{\text{injection acid}} = 4.17 \times 10^{-4} \text{ m/s}$	$T_{\text{injection acid}} = 320 \text{ K}$
<b>Wormhole</b>	$v_{\text{injection acid}} = 4.17 \times 10^{-5} \text{ m/s}$	$T_{\text{injection acid}} = 360 \text{ K}$

509  
 510 Overall, the results demonstrate that increasing the injection velocity leads to a transition from  
 511 wormhole-like to ramified and eventually uniform dissolution patterns. This behavior confirms the  
 512 earlier observation that raising acid temperature produces similar physical effects to lowering injection  
 513 velocity, as both alter the balance between reaction and transport processes. These insights are critical  
 514 for optimizing acidizing strategies, highlighting the importance of simultaneously managing  
 515 temperature and flow rate to achieve the desired dissolution morphology and stimulation efficiency.

516 **5. Conclusion and Future Work**

517 In this work, we developed a novel two-phase thermal Darcy–Brinkman–Forchheimer (DBF) model to  
 518 simulate matrix acidization processes in porous media under realistic reservoir conditions. The model  
 519 integrates multiphase flow, reactive transport, heat transfer, and dynamic porosity evolution into a  
 520 unified framework. Key contributions include the incorporation of temperature-dependent reaction  
 521 kinetics via Arrhenius-type relationships, as well as a fully coupled numerical solution strategy that  
 522 accounts for viscous, inertial, and thermal effects.

523 Numerical experiments were conducted to investigate the influence of initial matrix temperature,  
 524 injected acid temperature, and injection velocity on the acidizing process. Results showed that the  
 525 initial matrix temperature has minimal impact on wormhole formation and breakthrough behavior due  
 526 to rapid thermal equilibration driven by acid injection. In contrast, the injected acid temperature  
 527 significantly affects the dissolution pattern, reaction rate, and acid utilization efficiency. Higher  
 528 injection temperatures enhance reaction rates and promote localized wormhole formation, resulting in  
 529 earlier breakthrough and reduced pore volume to breakthrough (PVBT).

532 Furthermore, verification simulations demonstrated that increasing acid temperature has a similar effect  
533 to decreasing injection velocity. Both approaches shift the dissolution regime from uniform to  
534 wormhole-dominated patterns, highlighting a fundamental coupling between thermal and  
535 hydrodynamic controls. These insights offer practical guidance for optimizing acidizing strategies by  
536 balancing thermal input and flow rate to improve treatment efficiency and control wormhole geometry.  
537 While the current model provides a robust platform for simulating two-phase thermal acidizing, several  
538 directions remain for further enhancement. Future extensions will incorporate complex geological  
539 features such as natural fractures, vugs, and spatial heterogeneity in porosity and permeability. These  
540 additions will enable more realistic simulation of carbonate formations and fractured reservoirs.  
541 Moreover, the current study focuses on lab-scale domains. Scaling up the framework to field-scale  
542 models will require further optimization of the numerical solver, including parallel performance tuning  
543 and adaptive mesh refinement techniques. Besides that, incorporating multi-mineral reaction kinetics  
544 and competing acid-rock interactions (e.g., dolomite and calcite dissolution) will enhance the chemical  
545 fidelity of the model and broaden its applicability to various rock types. To better evaluate stimulation  
546 effectiveness, coupling the matrix acidizing model with wellbore dynamics and reservoir-scale  
547 production forecasting will be explored. Furthermore, data from high-fidelity simulations can be used  
548 to train surrogate models or integrate with machine learning algorithms for real-time optimization of  
549 injection schedules, temperature control, and acid composition.  
550 These directions will further strengthen the capability of the proposed framework for practical reservoir  
551 engineering applications and enable more accurate design and evaluation of acidizing treatments in  
552 complex subsurface environments.

#### 553 **Code availability**

554 Name of the code: Two-phase thermal DBF matrix acidizing simulator  
555 Contact: wuyuanq@gmail.com, +86-13928772190  
556 Hardware requirements: CPU: 2.2 GHz 6-Core Intel Core i7, Memory: 16 GB 2400 MHz DDR4  
557 Program language: FORTRAN90, MATLAB, TECPLOT  
558 Software required: LAPACK, UMFPACK, MUMPS, HYPRE, MPI  
559 Program size: more than 10000 lines  
560 The source codes are available for downloading at the link: <https://doi.org/10.5281/zenodo.17272888>  
561 (Wu, 2025)

#### 562 **Author contribution**

563 Wu: programming, the design and realization of the numerical experiments, the writing of the work.  
564 Kou: concepts of the work, the development of the numerical models, the algorithm design of the  
565 numerical scheme, the writing of the work.

566 **Competing interests**

567 The authors declare that they have no conflict of interest.

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571 final content.

572 **References**

- 573 Alokaily, S. A.: Numerical Simulations of Fluid Transport in Heterogeneous Darcy–Forchheimer–  
574 Brinkman Layers, *Journal of Fluids Engineering*, 144(12), 121404,  
575 <https://doi.org/10.1115/1.4055072>, 2022.
- 576 Araújo, E. D. A., Schwalbert, M. P., Leitão, R. J., & Aum, P. T. P.: Influence of Matrix-Acidizing  
577 Design on Oil Recovery and Economics in Carbonate Reservoirs Undergoing Waterflooding  
578 Offshore in Brazil, *Energies*, 17(4), 883, <https://doi.org/10.3390/en17040883>, 2024.
- 579 Babaei, M., & Sedighi, M.: Impact of phase saturation on wormhole formation in rock matrix acidizing,  
580 *Chemical Engineering Science*, 177, 39-52, <https://doi.org/10.1016/j.ces.2017.10.046>, 2018.
- 581 Daccord, G., Lenormand, R., & Lietard, O.: Chemical dissolution of a porous medium by a reactive  
582 fluid—I. Model for the “wormholing” phenomenon, *Chemical Engineering Science*, 48(1), 169-  
583 178, [https://doi.org/10.1016/0009-2509\(93\)80293-Y](https://doi.org/10.1016/0009-2509(93)80293-Y), 1993. a
- 584 Daccord, G., Lietard, O., & Lenormand, R.: Chemical dissolution of a porous medium by a reactive  
585 fluid—II. Convection vs reaction, behavior diagram, *Chemical engineering science*, 48(1), 179-  
586 186, [https://doi.org/10.1016/0009-2509\(93\)80294-Z](https://doi.org/10.1016/0009-2509(93)80294-Z), 1993. b
- 587 Deb, N., Farshi, M. S., Das, P. K., & Saha, S.: Convective flow optimization inside a lid-driven  
588 chamber with a rotating porous cylinder using Darcy–Brinkman–Forchheimer model, *Journal of*  
589 *Thermal Analysis and Calorimetry*, 149(12), 6125-6146, [https://doi.org/10.1007/s10973-024-](https://doi.org/10.1007/s10973-024-13228-y)  
590 13228-y, 2024.
- 591 Elsafih, M., & Fahes, M.: Quantifying the effect of multiphase flow on matrix acidizing in oil-bearing  
592 carbonate formations, *SPE Production & Operations*, 36(04), 795-806,  
593 <https://doi.org/10.2118/205397-PA>, 2021.
- 594 Fredd, C. N., & Fogler, H. S.: Influence of transport and reaction on wormhole formation in porous  
595 media, *AIChE journal*, 44(9), 1933-1949, <https://doi.org/10.1002/aic.690440902>, 1998.
- 596 Golfier, F., Zarcone, C., Bazin, B., Lenormand, R., Lasseux, D., & Quintard, M.: On the ability of a  
597 Darcy-scale model to capture wormhole formation during the dissolution of a porous medium,  
598 *Journal of fluid Mechanics*, 457, 213-254, <https://doi.org/10.1017/S0022112002007735>, 2002.
- 599 Jia, C., Sepehrmoori, K., & Yao, J.: Numerical studies and analysis on reactive flow in matrix acidizing  
600 coupled thermal-hydrological-mechanical-chemical processes, In *ARMA US Rock*  
601 *Mechanics/Geomechanics Symposium* (pp. ARMA-2021). ARMA, 2021 June.
- 602 Kou, J., Sun, S., & Wu, Y.: Mixed finite element-based fully conservative methods for simulating  
603 wormhole propagation, *Computer Methods in Applied Mechanics and Engineering*, 298, 279-302,  
604 <https://doi.org/10.1016/j.cma.2015.09.015>, 2016.

605 Kou, J., Sun, S., & Wu, Y.: A semi-analytic porosity evolution scheme for simulating wormhole  
606 propagation with the Darcy–Brinkman–Forchheimer model, *Journal of Computational and*  
607 *Applied Mathematics*, 348, 401-420, <https://doi.org/10.1016/j.cam.2018.08.055>, 2019.

608 Kou, J., Du, S., & Zhong, Z.: Energy stable modeling of two-phase flow in porous media with fluid–  
609 fluid friction force using a Maxwell-Stefan-Darcy approach, *Physics of Fluids*, 33, 073312,  
610 <https://doi.org/10.1063/5.0053373>, 2021.

611 Kou, J., Chen, H., Du, S., & Sun, S.: An efficient and physically consistent numerical method for the  
612 Maxwell-Stefan-Darcy model of two-phase flow in porous media, *International Journal for*  
613 *Numerical Methods in Engineering*, 124(3), 546-569, <https://doi.org/10.1002/nme.7131>, 2023.

614 Kou, J., Wang, X.: Numerical modeling of unsaturated flow in porous media using a thermodynamical  
615 approach, *Capillarity*, 11, 63-69, <https://doi.org/10.46690/capi.2024.06.01>, 2024.

616 Ma, G., Chen, Y., Wang, H., Li, T., & Nie, W.: Numerical analysis of two-phase acidizing in fractured  
617 carbonate rocks, *Journal of Natural Gas Science and Engineering*, 103, 104616,  
618 <https://doi.org/10.1016/j.jngse.2022.104616>, 2022.

619 Mahdavi Kalatehno, J., Khamehchi, E., Keihanikamal, M., Yousefmarzi, F., Dargi, M., & Daneshfar,  
620 P.: A successful case study of using HCl and viscoelastic diverting acid systems for carbonate  
621 matrix acidizing in an oil well with optimized predictive model, *Journal of Petroleum Exploration*  
622 *and Production Technology*, 15(1), 1, <https://doi.org/10.1007/s13202-024-01896-3>, 2025.

623 Panga, M. K., Ziauddin, M., & Balakotaiah, V.: Two - scale continuum model for simulation of  
624 wormholes in carbonate acidization, *AIChE journal*, 51(12), 3231-3248,  
625 <https://doi.org/10.1002/aic.10574>, 2005.

626 Qureshi, U., Qureshi, H. A., Bhatti, A. A., Saeed, M. S., & Khalid, M. S.: Investigating the effects of  
627 matrix acidizing and acid fracturing on the production optimization of a carbonate reservoir: a  
628 case study, *Arabian Journal of Geosciences*, 16(11), 620, [https://doi.org/10.1007/s12517-023-](https://doi.org/10.1007/s12517-023-11734-1)  
629 [11734-1](https://doi.org/10.1007/s12517-023-11734-1), 2023.

630 Sabooniha, E., Rokhforouz, M. R., Kazemi, A., & Ayatollahi, S.: Numerical analysis of two-phase  
631 flow in heterogeneous porous media during pre-flush stage of matrix acidizing: Optimization by  
632 response surface methodology, *Physics of Fluids*, 33(5), <https://doi.org/10.1063/5.0046106>, 2021.

633 Shahid, A., Wei, W., Abbas, T., & Bhatti, M. M.: A computational investigation of diffusivities and  
634 heat transfer in the flow of viscoelastic fluid through Darcy–Brinkman–Forchheimer medium,  
635 *Numerical Heat Transfer, Part B: Fundamentals*, 86(3), 762-779,  
636 <https://doi.org/10.1080/10407790.2023.2296076>, 2025.

637 Wei, W., Varavei, A., & Sepehrmoori, K.: Modeling and analysis on the effect of two-phase flow on  
638 wormhole propagation in carbonate acidizing, *SPE Journal*, 22(06), 2067-2083,  
639 <https://doi.org/10.2118/186111-pa>, 2017.

640 Wu, Y.: Parallel reservoir simulations with sparse grid techniques and applications to wormhole  
641 propagation, 2015.

642 Wu, Y. (2025). Two-phase thermal DBF matrix acidizing simulator. Zenodo.  
643 <https://doi.org/10.5281/zenodo.17272888>

644 Wu, Y., Kou, J., & Sun, S.: Matrix acidization in fractured porous media with the continuum fracture  
645 model and thermal Darcy-Brinkman-Forchheimer framework, *Journal of Petroleum Science and*  
646 *Engineering*, 211, 110210, <https://doi.org/10.1016/j.petrol.2022.110210>, 2022.

647 Wu, Y., Kou, J., Sun, S., & Wu, Y. S.: Thermodynamically consistent Darcy–Brinkman–Forchheimer  
648 framework in matrix acidization, *Oil & Gas Science and Technology–Revue d’IFP Energies*  
649 *nouvelles*, 76, 8, <https://doi.org/10.2516/ogst/2020091>, 2021.

- 650 Wu, Y., Salama, A., & Sun, S.: Parallel simulation of wormhole propagation with the Darcy–  
651 Brinkman–Forchheimer framework, *Computers and Geotechnics*, 69, 564-577,  
652 <https://doi.org/10.1016/j.compgeo.2015.06.021>, 2015.
- 653 Wu, Y., & Ye, M.: A Newton’s second law abided Darcy-Brinkman-Forchheimer framework in matrix  
654 acidization simulation, In *International Conference on Computational & Experimental*  
655 *Engineering and Sciences* (pp. 861-866), Cham: Springer International Publishing,  
656 [https://doi.org/10.1007/978-3-030-27053-7\\_73](https://doi.org/10.1007/978-3-030-27053-7_73), 2019 March.
- 657 Yoon, H. C., & Mallikarjunaiah, S. M.: A stabilized finite element method for steady Darcy-Brinkman-  
658 Forchheimer flow model with different viscous and inertial resistances in porous media, *arXiv*  
659 preprint [arXiv:2501.04041](https://arxiv.org/abs/2501.04041), 2025.
- 660 Zhou, L., Guo, A., Wang, X., Qiao, J., & Tang, X.: The effect of temperature, natural fractures and  
661 vugs on the acidizing process in fractured-vuggy reservoirs with hydro-thermal-chemical coupled  
662 modeling, *Journal of Petroleum Science and Engineering*, 213, 110416,  
663 <https://doi.org/10.1016/j.petrol.2022.110416>, 2022.

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