

Relative Permeability Variation Depending on Viscosity Ratio and Capillary Number

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November 30, 2022

Abstract

The relative roles of parameters governing relative permeability, a crucial property for two-phase fluid flows, are incompletely known. To characterize the influence of viscosity ratio (M) and capillary number (Ca), we calculated relative permeabilities of nonwetting fluids (knw) and wetting fluids (kw) in a 3D model of Berea sandstone under steady-state condition using the lattice Boltzmann method. We show that knw increases and kw decreases as M increases due to the lubricating effect, locally occurred pore-filling behavior, and instability at fluid interfaces. We also show that knw decreases markedly at low Ca ($\log Ca < -1.25$), whereas kw undergoes negligible change with changing Ca . An M - Ca - knw correlation diagram, displaying the simultaneous effects of M and Ca , shows that they cause knw to vary by an order of magnitude. The color map produced is useful to provide accurate estimates of knw in reservoir-scale simulations and to help identify the optimum properties of the immiscible fluids to be used in a geologic reservoir.

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Key Points:

- Relative permeability in two-phase flow is calculated in a three-dimensional digital Berea rock using Lattice Boltzmann Method
- Relative permeability varies due to lubrication effect, shear force and capillary force, and is related to fluid droplet fragmentation
- Relative permeability on viscosity ratio-capillary number map is created to predict spatiotemporal variation of reservoir permeability

Abstract

The relative roles of parameters governing relative permeability, a crucial property for two-phase fluid flows, are incompletely known. To characterize the influence of viscosity ratio (M) and capillary number (Ca), we calculated relative permeabilities of nonwetting fluids (k_{nw}) and wetting fluids (k_w) in a 3D model of Berea sandstone under steady-state condition using the lattice Boltzmann method. We show that k_{nw} increases and k_w decreases as M increases due to the lubricating effect, locally occurred pore-filling behavior, and instability at fluid interfaces. We also show that k_{nw} decreases markedly at low Ca ($\log Ca < -1.25$), whereas k_w undergoes negligible change with changing Ca . An M - Ca - k_{nw} correlation diagram, displaying the simultaneous effects of M and Ca , shows that they cause k_{nw} to vary by an order of magnitude. The color map produced is useful to provide accurate estimates of k_{nw} in reservoir-scale simulations and to help identify the optimum properties of the immiscible fluids to be used in a geologic reservoir.

Plain Language Summary

The relative permeability is a crucial parameter in a system where two fluid phases exist simultaneously. For example, in carbon capture and storage, relative permeability is important to assess the replacement mechanism of the existing fluid in the reservoir (wetting fluid) by the injected CO₂ (non-wetting fluid). It is also an important parameter in enhanced oil recovery fields, as high relative permeability of oil indicates that the oil in the reservoir can be extracted quickly. The relative permeability is temporally and spatially varied by reservoir conditions (e.g., temperature). But currently, in reservoir-scale fluid flow simulation, relative permeability is assumed to be constant regardless of the different conditions. In this study, we conducted

simulations to calculate relative permeability in various viscosity ratio (M) and capillary number (Ca) conditions. We found that relative permeability changes dramatically in different M and Ca conditions, and we further mapped relative permeability on the diagram between M and Ca to predict relative permeability accurately in various reservoir conditions. Our findings can be useful to determine the suitable fluid properties to be used in reservoir management and to accurately estimate fluid behavior based on reservoir-scale simulation with variant relative permeability.

1 Introduction

The relative permeability of the different fluids in a two-phase flow has been extensively studied in scientific and engineering fields concerned with two-phase flows in geological reservoirs, such as in carbon capture and storage (CCS) fields, enhanced oil recovery (EOR) operations, geothermal power systems, and geological radioactive waste disposal repositories (Benson et al., 2015; C. Chen & Zhang, 2010; Gudjonsdottir et al., 2015; Niibori et al., 2011; Shad et al., 2008; Wu & Wang, 2020). Relative permeability is a crucial hydraulic property for modeling the flow of both fluids and assessing the mechanisms of fluid displacement in the reservoir. For example, when CO_2 is injected into a CCS reservoir, the CO_2 (nonwetting phase) displaces the existing fluid in the reservoir, such as oil or brine (wetting phase). Relative permeability values can be used to estimate the reduction in CO_2 fluid flow due to surface-tension effects between CO_2 and the brine, thus the parameter is useful to assess the injectivity of the CO_2 (Benson et al., 2015; Burnside & Naylor, 2014). The relative permeability can also aid estimation of how much fluid can be displaced by CO_2 before the system reaches the wetting fluid irreducible saturation condition, limiting the CO_2 volume that can be stored in the reservoir

(Burnside & Naylor, 2014). Conversely, in EOR systems, the best results are obtained when the relative permeability is high for oil and low for the injected fluid (Heins et al., 2014).

Several factors influence relative permeability in a two-phase flow system. For example, in a CCS project, the fluid relative permeability is affected by the heterogeneity of the rock, such as pore size and pore connectivity (Benson et al., 2015; Fei Jiang & Tsuji, 2014; Zhang et al., 2022). In addition, relative permeability is affected by the interactions between the two fluids, such as interfacial tension, the viscosity and velocity of the fluids, and their wettability (Lefebvre du Prey, 1973). Because fluid viscosity varies dramatically with pressure and temperature, such reservoir environments can further affect relative permeability. Therefore, in a two-phase flow system, the relative permeability value of each fluid is not only a function of saturation; it is also affected by other parameters related to environmental and by the interaction between the two component fluids.

Lenormand et al. (1988) reported that two parameters, viscosity ratio and capillary number, can explain the interaction between two immiscible fluids. The viscosity ratio (M) is a dimensionless parameter describing the ratio between the viscosity of the injected nonwetting fluid and the viscosity of the ambient wetting fluid:

$$M = \frac{\mu_{nw}}{\mu_w} \quad (1)$$

where μ_{nw} is the dynamic viscosity of the nonwetting fluid and μ_w is the dynamic viscosity of the wetting fluid.

The capillary number (Ca) is a dimensionless parameter describing the ratio between the viscous drag forces and the interfacial tension forces between two immiscible fluids:

$$Ca = \frac{\mu_{nw} V_{nw}}{\sigma} \quad (2)$$

where V_{nw} is the average fluid velocity of the nonwetting fluid and σ is the interfacial tension (IFT) between the two fluids.

Despite the recognition that the relative permeability k is a function of M , several studies of this relationship have reported divergent results. An experimental study (Odeh, 1959) found that the relative permeability of the nonwetting fluid (k_{nw}) increases and the relative permeability of the wetting fluid (k_w) stays relatively constant as M increases, as have several other experimental and numerical studies (Dou & Zhou, 2013; Goldsmith & Mason, 1963; Huang & Lu, 2009; Jeong et al., 2017; Mahmoudi et al., 2017; Yiotis et al., 2007; Zhao et al., 2017). An analytical study of co-current annular flow in which the wetting fluid is distributed on the pore surface and the nonwetting fluid is in the middle of the pore produced empirical equations for the nonwetting and wetting fluids as a function of M and saturation:

$$k_{nw} = S_{nw} \left[\frac{3}{2} M + S_{nw}^2 \left(1 - \frac{3}{2} M \right) \right] \quad (3)$$

$$k_w = \frac{1}{2} (1 - S_w)^2 (3 - S_w) \quad (4)$$

where S_{nw} is the saturation of the nonwetting fluid, and $S_w = 1 - S_{nw}$ is the saturation of the wetting fluid. These equations suggest that k_w is not affected by increasing M and is a function of saturation alone. However, other studies have reported that k_{nw} increases and k_w decreases as M increases (Ahmadlouydarab et al., 2012; Fan et al., 2019; Goel et al., 2016; Ramstad et al., 2010); thus, there is as yet no general agreement on the variation of k_w with increasing M . One of the challenges of previous studies was the difficulty of removing the effects of capillary forces and wettability factors when evaluating this relationship.

Similarly, studies of the influence of Ca on relative permeability is still incompletely known. In an experimental study, Fulcher et al. (1985) concluded that k_{nw} is a function of IFT and viscosity variables individually rather than a function of the Ca , whereas k_w can be modeled directly as a function of Ca . Several studies (Asar & Handy, 1989; Fan et al., 2019; Harbert, 1983) also found that both k_{nw} and k_w increase as IFT decreases because the two fluids interfere less with each other and thus tend to form more well-connected flow pathways. One of the studies (Asar & Handy, 1989) showed that both fluids relative permeability curves tend to straighten and approach the 45° tangent line as IFT approaches zero. Other studies (Amaefule & Handy, 1982; Fei Jiang et al., 2014; Shen et al., 2010) also concluded that both k_{nw} and k_w decrease as Ca decreases. On the other hand, a numerical study (Zhao et al., 2017) concluded that k_w increases with increasing Ca under neutral wetting conditions ($\theta = 90^\circ$) but stays relatively constant with increasing Ca under strong wetting conditions ($\theta = 135^\circ$). These divergent results warrant further investigations of how relative permeability changes with changing IFT and Ca . One of the challenges in this evaluation is the difficulty of isolating the effect of IFT while keeping other parameters, such as viscosity, constant. In addition, Ca is a function of fluid velocity, which is a direct of from the simulation. Thus, it is difficult to hold Ca constant in all simulation conditions.

Previous studies have also demonstrated how the viscosity ratio can affect the breakup of fluid droplets. Instabilities due to the viscosity gradient at the interface of the two fluids cause fluid droplet deformation and breakup under viscous flow conditions when the Reynolds number is low (Bischofberger et al., 2015; Mu'min et al., 2021; Nekouei & Vanapalli, 2017; Stone, 1994). We suspect that the fluid droplet fragmentation mechanism is also one of the driving factors of the relative permeability change caused by the viscosity ratio. However, to the best of

our knowledge, no research linking relative permeability variation to fluid droplet breakup caused by viscosity ratio variation has ever been done.

Previous studies have usually evaluated the separate effects of M and Ca on relative permeability. However, in two-phase flows, the effects of viscosity gradient and interfacial tension must be evaluated simultaneously to accurately predict the hydraulic properties in the system (Tsuji et al., 2016), including relative permeability. Because few studies have evaluated the condition when both parameters influence relative permeability using a steady-state simulation, our aim in this work was to fill that knowledge gap.

In this paper, we propose a method to simultaneously evaluate the effects of viscosity ratio M and capillary number Ca on relative permeability by creating a M – Ca – k_{nw} relationship map (i.e., showing relative permeability on a diagram relating M and Ca) by applying a color gradient Lattice Boltzmann method (LBM) simulation to a three-dimensional (3D) digital rock model. We begin by evaluating the effects of M and Ca individually on the relative permeabilities of the nonwetting and wetting fluids by holding other parameters constant. We also calculate the number of fluid clusters to describe the fluid connectedness, a factor that is directly related to relative permeability, and link the fluid connectedness to fluid droplet breakup. We then map k_{nw} for various M – Ca conditions in search of general trends. We believe that such a map has never before been created, and that it will be useful for quickly estimating relative permeability if M and Ca are known. By understanding the variations of relative permeability in the M – Ca parameter space, we can conduct accurate large-scale reservoir simulations by considering the reservoir conditions M and Ca and thereby contribute to the wide range of research regarding applications that employ two-phase fluid mixtures.

2 Methods

2.1 Lattice Boltzmann Method

The LBM is a branch of computational fluid dynamics that has emerged as a popular technique to solve multiphase fluid flow systems in complex geometries because of its algorithmic simplicity (S. Chen & Doolen, 1998; Dou & Zhou, 2013; Fei Jiang et al., 2014; Ramstad et al., 2010; Succi et al., 2010). LBM simulations treat fluids as a group consisting of fictive particles; the movement of these particles is simulated with a statistical approach. The movement of the bulk fluid is simulated from the propagation and collision processes of the fictive particles (Huang et al., 2011). We chose the Rothman–Keller color gradient model (Tölke et al., 2006) to conduct the simulations because it can simulate a high fluid viscosity ratio with better accuracy than other LBM models (Ahrenholz et al., 2008; Yang & Boek, 2013). The color gradient model is so named because it graphically represents two-phase fluids as a mixture of a wetting and nonwetting fluid, assigned the colors blue and red, respectively (Huang et al., 2015; Fei Jiang et al., 2022). Our 3D simulations used the D3Q19 (3 dimensions, 19 velocity lattice) velocity model.

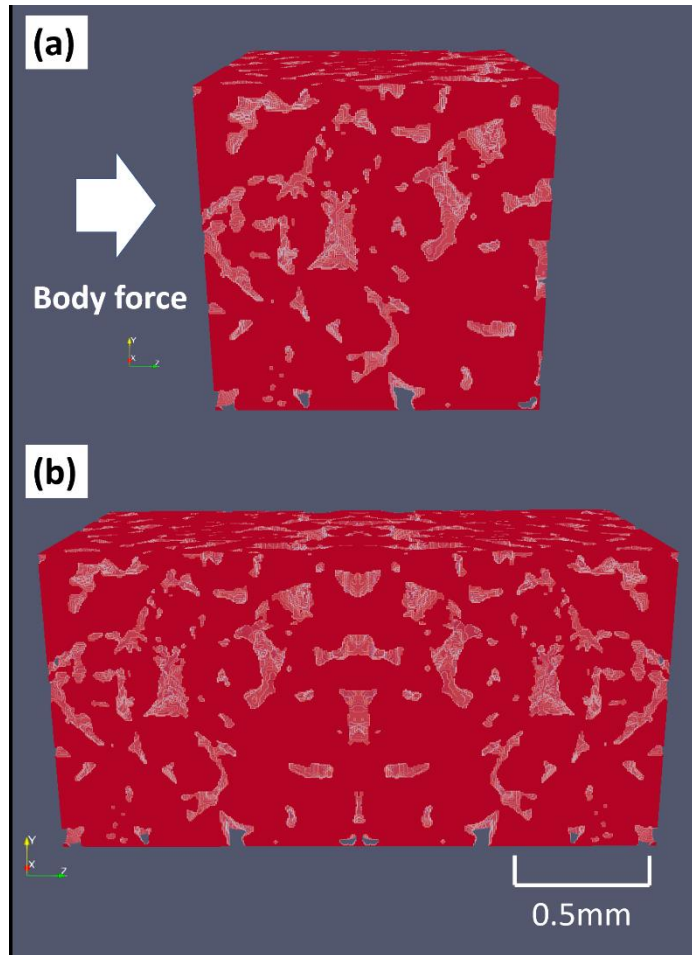


Fig. 1. Berea sandstone digital rock model: (a) core form and (b) mirrored form.

2.2 Digital rock model

We used a 3D digital rock model obtained from microtomographic images of Berea sandstone (Fig. 1a). Berea sandstone is chosen due to its relatively well-known properties (Øren & Bakke, 2003). The rock has a relatively large mean grain size of $\sim 250 \mu\text{m}$ and consists of quartz, feldspar, carbonates, and clay minerals. Its average pore size is $\sim 20 \mu\text{m}$ with an average throat size of $\sim 10 \mu\text{m}$. Our rock model, obtained from the database of Dong and Blunt (2009), had an original resolution of $5.345 \mu\text{m}$ per pixel. To ensure that the digital rock resolution would be fine enough for the two-phase flow simulation conditions, each pixel was split into 2; hence,

each pixel in our model represents 2.673 μm . The 3D digital rock is a stack of 400 tomographic images, and each image includes 400×400 pixels. Thus, the digital rock represents an actual size of 1.069 mm $\times$ 1.069 mm $\times$ 1.069 mm. The simulation requires a rock model that is larger than its representative elementary volume (~ 1 mm for Berea sandstone), and the multi-pore nature of the rock must be considered.

2.3 Relative permeability calculation

The viscosity ratio M is defined by dividing the nonwetting fluid's viscosity (μ_{nw}) by the wetting fluid's viscosity (μ_{w}) (equation 1). In the LBM, μ can be calculated as follows:

$$\mu = \frac{1}{3} \left(\tau - \frac{1}{2} \right) \quad (5)$$

where τ represents the relaxation parameter. Thus, the viscosities of both fluids can be modified by changing the τ value. Because τ must be larger than 0.5 for a positive viscosity and the simulation becomes unstable as τ approaches 0.5, it is important that the simulation uses a τ value that ensures its stability and accuracy. In this study, we altered M only by changing the τ value of the nonwetting fluid to manipulate μ_{nw} . The wetting fluid viscosity was kept constant at 0.1555 lu²/ts (length²/time in lattice units) in all simulations to preserve its stability. Changes in M can be interpreted as a change in either μ_{w} or μ_{nw} ; thus, modifying either viscosity value will produce the same results if their ratio is maintained.

The capillary number Ca can be modified based on equation (2) by changing the σ and μ_{nw} parameters. Because μ_{nw} is also used to determine M , we achieved the desired Ca by adjusting σ alone. In the color gradient LBM model, σ can be modeled by introducing a perturbation term into equilibrium equations based on the gradient of the phase field of the two

fluids. Tölke et al. (2006) provides a detailed description of the model. The average velocity of the nonwetting fluid after the simulation had converged to equilibrium was used to calculate the Ca number. We adjusted σ at every simulation point (every M and in every saturation condition) to ensure that Ca had a similar value and fell within the permitted error range of $\log Ca \pm 0.10$. When the fluid velocity change in the last 1,000 simulation steps was less than 2% for all cases, the simulations were assumed to have converged at that time. The time-averaged velocity value in the last 3,000 steps was calculated for further investigation.

To remove the complications arising from wettability effects, we assumed that the solid was completely nonwetting ($\theta = 180^\circ$) to the nonwetting fluid in all conditions. Given that condition, the wetting fluid coats the surface of the rock, whereas the nonwetting fluid is not in contact with the pore wall and can only occupy the central parts of the pores.

We applied a steady-state simulation condition, which assumes that the wetting fluid saturation (S_w) and nonwetting fluid saturation (S_{nw}) are kept constant. In the initial state, both fluids were specified as being randomly distributed in the pore spaces. The initial condition was chosen because it allows each pore body to have roughly the same proportions of nonwetting and wetting fluid according to the prescribed saturation; thus, this condition produces an extremely uniform distribution of nonwetting and wetting fluid throughout the rock. In addition, because the initial fluid connectivity is reduced under this initial condition and increases capillary trapping, this initial distribution might be desirable in a CCS site (Fei Jiang & Tsuji, 2016). Under this initial condition, fluids are generated randomly in pore space; therefore, some nonwetting fluids touch the grain surface at the initial stage. As the simulation progresses, the wetting fluid coats the surface of the rock, and the nonwetting fluid moves to occupy only the central parts of pores because of the perfect wettability condition of the solid. A constant body

force was applied in the z direction to both fluids to mimic the pressure gradient: $g = dP/dz$ (Fig. 1a). For the interaction between rock solid nodes and fluid voxels, no-slip boundary conditions were applied by using the halfway bounce-back scheme (X. Li et al., 2016; Singh et al., 2017). The density of both fluids was set at 1.0 in lattice units, which corresponds to $1,000 \text{ kg/m}^3$ in the physical unit. We did this because our study was focused on viscosity differences, and the effect of density contrast is minor if inertial force can be neglected.

We applied a periodic boundary condition in the x , y , and z directions of our 3D model. Specifically, in the flow direction (z direction), the rock was mirrored to ensure that the pore spaces on the right side were connected to the left side of the digital rock (Fig. 1b) (F Jiang & Tsuji, 2017).

2.3.1 Relative permeability curve

The relative permeability curve is a plot of k versus saturation S_{nw} . As S_{nw} increases, k_{nw} increases and k_w decreases. The shape of this curve can change as a result of viscous drag force and capillary force; thus, an investigation of its variation with changes in M and Ca is important to determine the optimum conditions to achieve the desired value of k . We plotted k_{nw} and k_w as a function of S_{nw} to create relative permeability curves for various conditions.

In this study, we plotted all permeability curves for S_{nw} values within a wide range from 10% to 90%. Although some of these S_{nw} values might be unrealistic in actual injection settings because of irreducible wetting phase saturation (Tsuji et al., 2016), we still considered those hypothetical S_{nw} conditions for the ideal situation of a steady-state laboratory experiment.

2.3.2 M - Ca - k_{nw} permeability color map

After confirming the influence of M and Ca on k_{nw} and k_w in a two-phase flow system, we conducted simulations at various M and Ca conditions with S_{nw} held at a constant value of 20% to create a plot of M – Ca – k_{nw} correlation or color diagram. The 20% S_{nw} condition was chosen because it can realistically be achieved under all M – Ca conditions (Tsuji et al., 2016). The color diagram is useful to evaluate the degree of influence of M and Ca on changes of relative permeability when both parameters influence a system, and can also provide estimates of relative permeability in a reservoir under a wide range of M and Ca .

3 Results and interpretation

Before conducting simulations in various M and Ca conditions, we first ran single-phase simulations to determine the absolute permeability value of the porous media and to verify the stability of the simulation. The results are presented in Appendix A. Afterwards, the effects of M and Ca were then evaluated individually to see how they influence nonwetting and wetting relative permeability curves in a two-phase flow system. Next, we conducted simulations over a wide range of M and Ca values at a S_{nw} value of 20% to create a relative permeability color diagram.

3.1 Relative permeability curve from a two-phase simulation

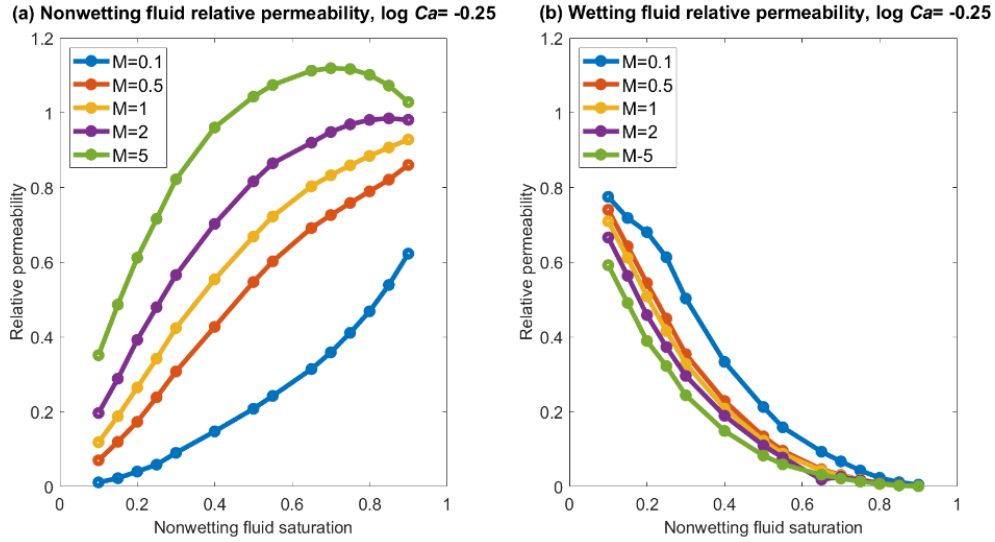


Fig. 2. Relative permeability curves as a function of M at constant $\log Ca = -0.25 \pm 0.1$ for $M = 0.1$ – 5 : (a) nonwetting fluid relative permeability (k_{nw}) curve and (b) wetting fluid relative permeability (k_w) curve.

3.1.1 Effect of changes in M on k_{nw} and k_w

We plotted values of k_{nw} for five values of M (0.1, 0.5, 1, 2, and 5) over the range of S_{nw} (10% to 90%) while holding $\log Ca$ nearly constant at -0.25 ± 0.10 (Fig. 2a). We evaluated the two-phase flow at high Ca to completely remove the capillary force effect. In all simulation conditions, the Reynolds number must not exceed 10 (Bear, 1975). The maximum Reynolds number reached in this set of simulations was 5.27 and 9.50 for nonwetting and wetting fluid, respectively. The M values (0.1, 0.5, 1, 2, and 5) were chosen to represent a wide range of mobility conditions, from an extremely unfavorable to an extremely favorable mobility condition. The k_{nw} value increased as M increased, especially for intermediate S_{nw} values (30% to 70%). For $M = 5$, k_{nw} exceeded 1 under some S_{nw} conditions; that is, it exceeded the intrinsic permeability of the rock. This result agrees with previous studies and can be explained by the

279 lubrication effect, also known as the viscous coupling effect (Goel et al., 2016; Fei Jiang et al.,
 280 2021; H. Li et al., 2005; Ramstad et al., 2010; Yiotis et al., 2007; Zhao et al., 2017).

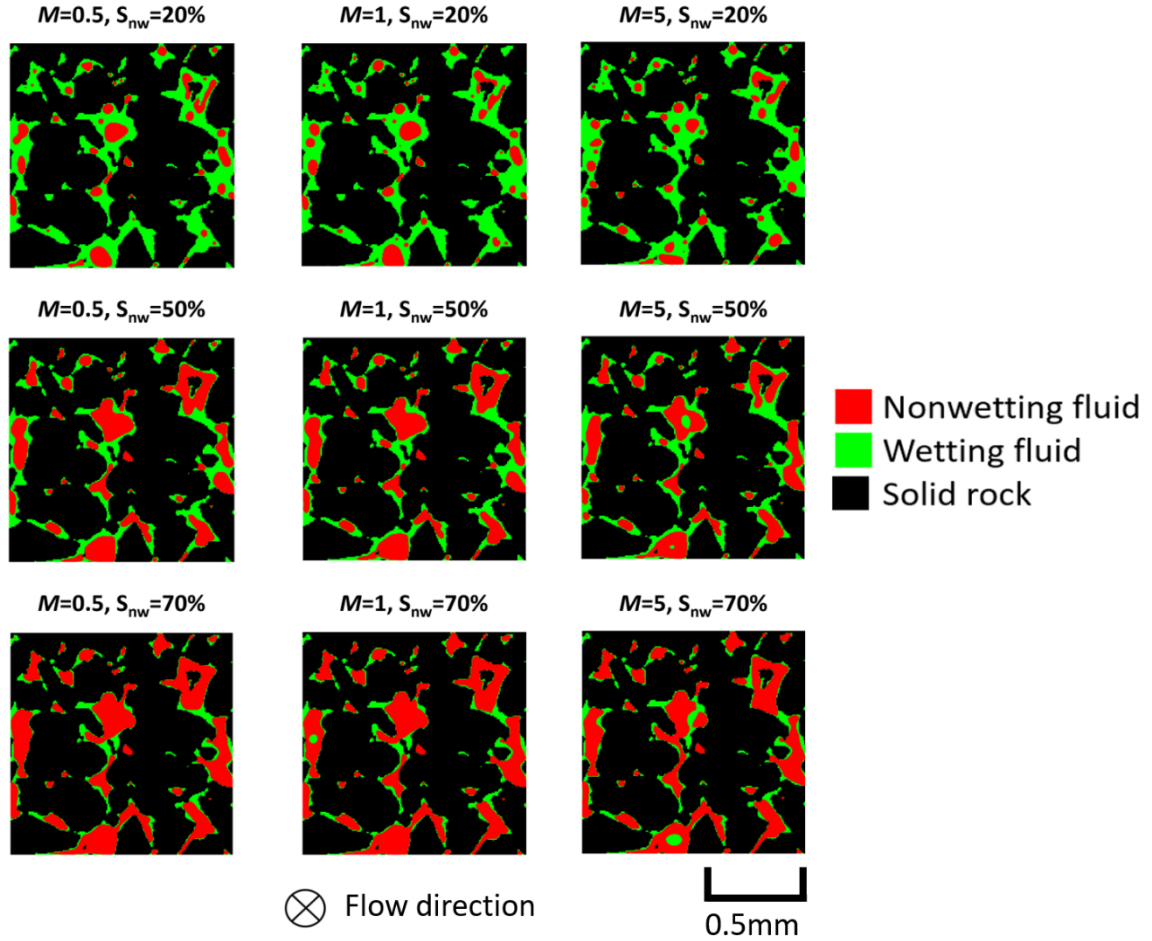


Fig. 3. Two-dimensional slice showing the distributions of nonwetting fluid (red), wetting fluid (green), and solid rock (black) at various values of M and S_{nw} .

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To illustrate the lubrication effect, we provide a 2D image showing the distribution of nonwetting fluid, wetting fluid, and solid rock at the final state of the simulation under various M and S_{nw} conditions (Fig. 3). Because the wetting fluid (green) layer is generally present between the nonwetting fluid and the rock, the wetting fluid moves along the rock surface and the nonwetting phase is confined to the central part of the pore. Thus, the velocity of the nonwetting

fluid is affected only by the momentum transfer across fluid–fluid interfaces and not by contact with the grain surface.

At sufficiently high values of M , the wetting fluid acts as a lubricant that enhances the movement of the nonwetting fluid, making its permeability higher than in the single-phase condition such that k_{nw} is greater than 1 (Fig. 2a). This lubrication effect most strongly affects k_{nw} in the intermediate S_{nw} range, when the wetting fluid forms thick films that provide the nonwetting fluid with a moving boundary (Vafai, 2000). At low S_{nw} , the nonwetting phase tends to form droplets with low connectivity; thus, the lubrication effect is less pronounced. At high S_{nw} , k_{nw} also decreases toward unity because less wetting fluid is present in the rock and the nonwetting phase comes into contact with the rock, degrading the lubrication effect and reducing k_{nw} near to the single-phase condition. Because the lubrication effect occurs at high M values, it is commonly observed in mixtures of heavy oil and water in EOR systems and has been confirmed by experimental results (Goel et al., 2016; Shad et al., 2008).

In addition to the lubrication effect, the relative permeability change due to the viscosity ratio can be linked to local pore-filling behavior. Because of the body force, a nonwetting fluid droplet can move from one pore body to neighboring pore bodies. As M increases, as a result of favorable mobility conditions, there is more chance for nonwetting fluid droplets to move to several neighboring pore bodies through the narrow pore throats (Bakhshian et al., 2021), causing fluid droplet fragmentation and thereby enabling the fluid to occupy smaller pore spaces. This might be one of the contributing factors causing k_{nw} to exceed 1 when $M = 5$ at intermediate saturation. Furthermore, under high S_{nw} , nonwetting fluid already exists in most of the pore bodies throughout the simulation time; thus, locally occurring pore-filling behavior effects decrease, causing k_{nw} to decrease towards unity under high S_{nw} conditions.

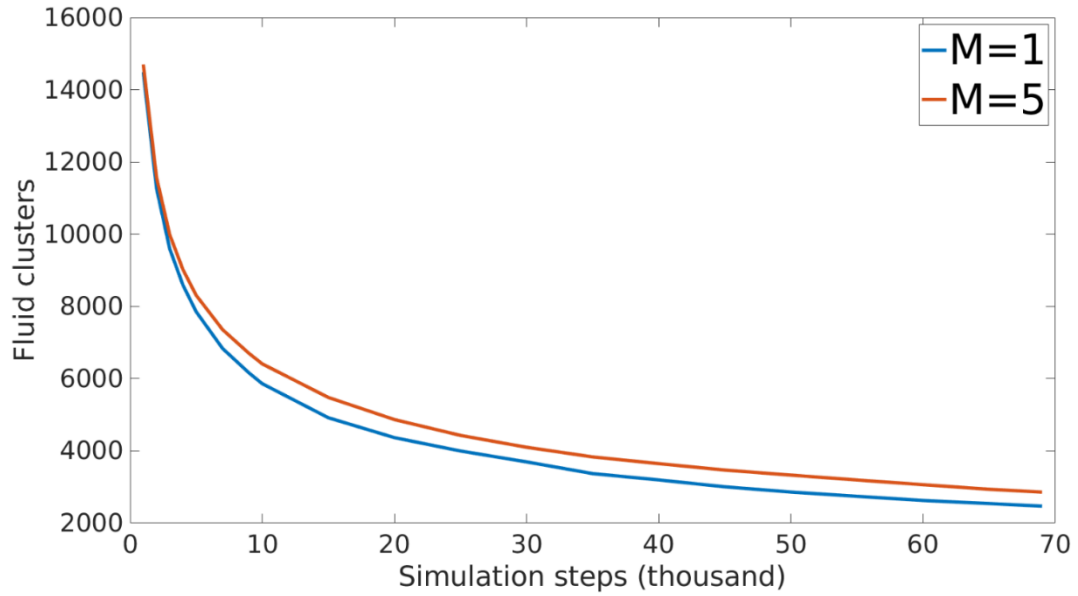


Fig. 4. Evolution of the number of clusters of nonwetting fluid at $M = 1$ and $M = 5$, with $S_{nw} = 20\%$ and $\log Ca = -0.25 \pm 0.1$.

To investigate nonwetting fluid droplet fragmentation throughout the simulation, we considered the evolution of the number of nonwetting fluid clusters during simulations with $M = 1$ and $M = 5$, $\log Ca = -0.25 \pm 0.10$, and $S_{nw} = 20\%$ (Fig. 4). Throughout the simulations, the number of nonwetting clusters for both simulation conditions decreases, suggesting that some of the nonwetting fluid clusters become more connected and form larger clusters as the simulation proceeds. In the final converged condition, the number of nonwetting fluid clusters is higher for $M = 5$ than for $M = 1$. This result indicates that the nonwetting fluid at $M = 5$ forms more individual clusters and has less fluid connectivity than at $M = 1$. This difference is the result of the instability of the interface of the fluids due to the viscosity stratification (Yiantsios & Higgins, 1988; Yih, 1967). When two immiscible fluids have different viscosities, their velocities will also be different at their interface, causing instability. The instability causes the nonwetting fluid to create more clusters. This result agrees with past research and our hypothesis that instability at the fluid interface can cause fluid droplet fragmentation (Bischofberger et al.,

2015; Mu'min et al., 2021; Nekouei & Vanapalli, 2017; Stone, 1994). As more fluid clusters are formed, the size of the nonwetting fluid clusters is smaller, and it is easier for them to pass through the pore space. At $M = 1$, the nonwetting fluid is more connected and its fluid clusters are larger. The larger clusters cannot pass through the small pore throats, leading to lower k_{nw} .

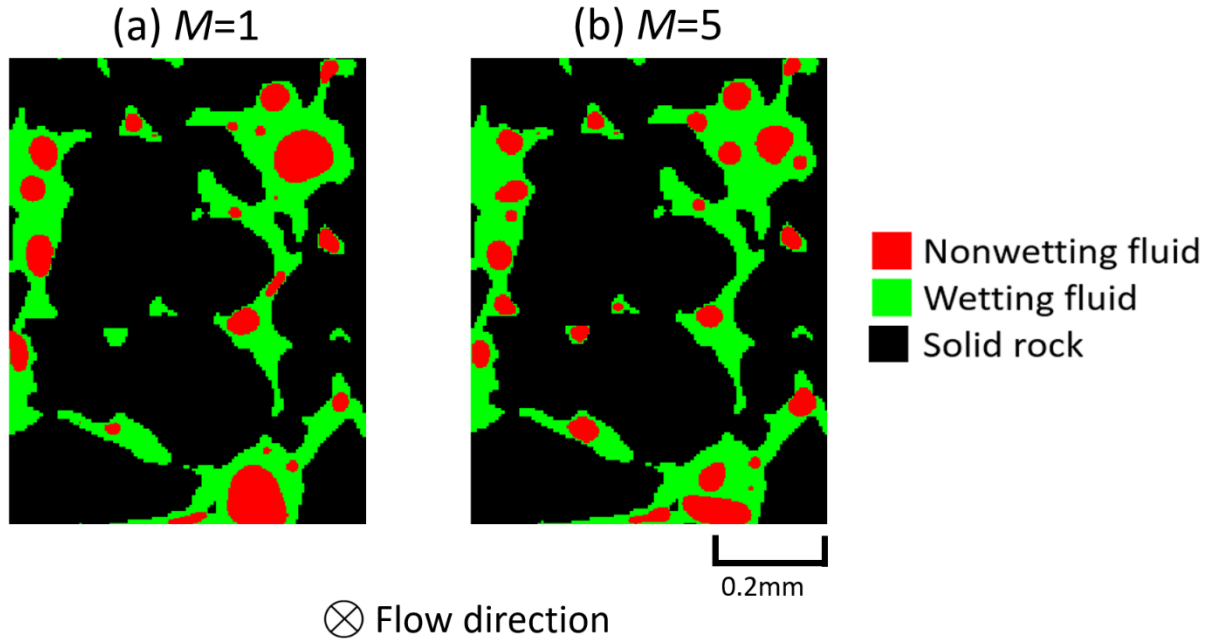


Fig. 5. Two-dimensional slice showing the distributions of nonwetting (red), wetting fluid (green), and solid rock (black) at (a) $M = 1$ and (b) $M = 5$ at 20% S_{nw} . These images show a portion of the slice illustrated in Fig. 3.

To obtain further insights into fluid cluster size, we provide an illustration showing nonwetting and wetting fluid distributions in an enlarged section of a 2D slice of the simulated specimen for $M = 1$ and $M = 5$ at $\log Ca = -0.25 \pm 0.10$ and 20% saturation after the simulation had converged (Fig. 5). The nonwetting fluid (red) at $M = 1$ forms larger clusters in the pore throat, whereas at $M = 5$, the nonwetting fluid forms a larger number of clusters but the size of each cluster is smaller. This result is comparable to our findings (shown in Fig. 4) that nonwetting fluid droplet fragmentation occurs as a result of instability at the fluid interface as M increases. Because the nonwetting fluid clusters are smaller, it is easier for them to pass through

the pore space, leading to higher k_{nw} . Furthermore, the wetting fluid (green) coats the grain surface under both conditions and produces a lubrication effect (Fig. 5).

Our results also demonstrate that k_w decreases as M increases (Fig. 2b), mainly when nonwetting fluid saturation is low. This phenomenon can be attributed to an increase of shear force from the nonwetting phase. As M increases and nonwetting fluid droplet fragmentation occurs, the wetting fluid has a larger contact area with the nonwetting fluid (Fig. 5). As a result, the wetting fluid receives a higher shear force from the nonwetting fluid, which inhibits the flow of the wetting fluid so that it has a lower velocity. In addition, as wetting fluid saturation decreases, the k_w gaps under all M conditions become smaller.

To confirm our interpretation, we calculated the degree of contact between nonwetting and wetting fluid by calculating the number of wetting fluid voxels that were adjacent to a nonwetting fluid voxel, for conditions of $M = 1$ and $M = 5$ at $\log Ca = -0.25 \pm 0.10$ and $S_{nw} = 20\%$. When the simulations converged, 13.77679% and 13.88825% of wetting fluid voxels were adjacent to nonwetting fluid voxels at $M = 1$ and $M = 5$, respectively. This result shows that, at $M = 5$, the wetting phase fluid has a higher degree of contact with the nonwetting fluid and receives more drag force from the nonwetting fluid; thus, k_w is lower at $M = 5$ than at $M = 1$, as shown in Fig. 2b.

In addition, Zhao et al. (2017) suggested that the k_w decrease in response to an M increase occurs because the wetting fluid is more attached to the grain surface when the viscosity ratio is large. When the viscosity ratio is high, fragmentation of the nonwetting phase tends to occur easily and the fragments move towards the centers of pore bodies, so that wetting fluid can easily move to coat the grain surface. Consequently, because the wetting fluid has a larger contact area with the grain surface, it receives more shear drag force, which causes a decrease in velocity. To

confirm our interpretation, we calculated the degree of contact between the wetting fluid and the grain surface, as indicated by the number of wetting fluid voxels having a solid surface as a neighbor, for $M = 1$ and $M = 5$ at $\log Ca = -0.25 \pm 0.10$ and $S_{nw} = 20\%–70\%$ after the simulation had converged (Table 1).

	$M=1$	$M=5$
$S_{nw}=20\%$	36.79546%	36.74534%
$S_{nw}=50\%$	50.94340%	51.18913%
$S_{nw}=70\%$	67.81628%	68.25865%

Table 1. Percentage of wetting fluid that has a grain surface as a neighbor voxel fluid at $M = 1$ and $M = 5$, with S_{nw} of 20%–70% and $\log Ca = -0.25 \pm 0.1$, after the simulation has converged.

At $S_{nw} = 20\%$, the percentage of wetting fluid voxels adjacent to a solid surface is low, because the wetting fluid film between the nonwetting fluid and grain surface is thick (Fig. 3) and only a small portion of the wetting fluid is in direct contact with the grain surface. The values for $M = 1$ and $M = 5$ are also similar, suggesting that the nonwetting fluid at low S_{nw} has less chance to approach the rock surface because of the thick wetting fluid film even as M increases. Thus, at low S_{nw} , the dominant cause of the k_w decrease is the shear force at the fluid interface and not the shear drag force at the wall. When $S_{nw} = 50\%$ and 70% and $M = 5$, the proportion of the wetting phase fluid attached to the grain surface increases compared to when $M = 1$, indicating that the wetting fluid has a higher degree of contact and receives more drag force from the solid wall. Thus, k_w at $M = 5$ is lower than that at $M = 1$ (Fig. 2b).

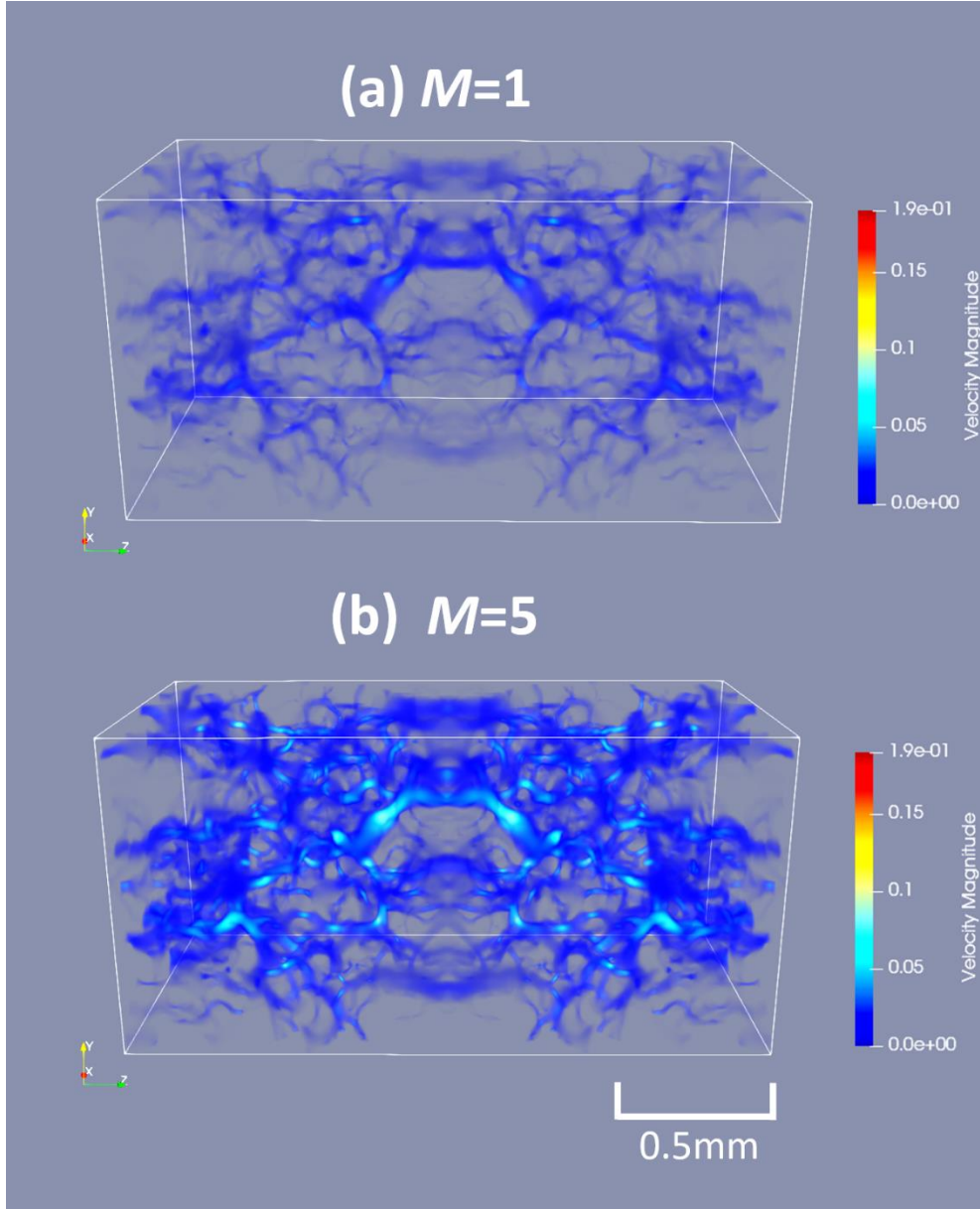


Fig. 6. Normalized velocity fields for the nonwetting fluid in a two-phase simulation with (a) $M = 1$ and (b) $M = 5$, at $\log Ca = -0.25 \pm 0.1$ and $S_{nw} = 20\%$.

The results of fluid velocity for $M=1$ and $M=5$ in the pore spaces at 20% saturation after the simulation converges are plotted in Figs. 6a and 6b, respectively. The fluid velocity field in Fig. 6 is normalized to consider the effect of viscosity change using the equation $U^* = \frac{U}{\Delta PL/\mu}$. It

can be seen that the normalized fluid velocity is higher at $M = 5$ than at $M = 1$, indicating that k_{nw} is higher at $M = 5$ than at $M = 1$, in agreement with our relative permeability curve.

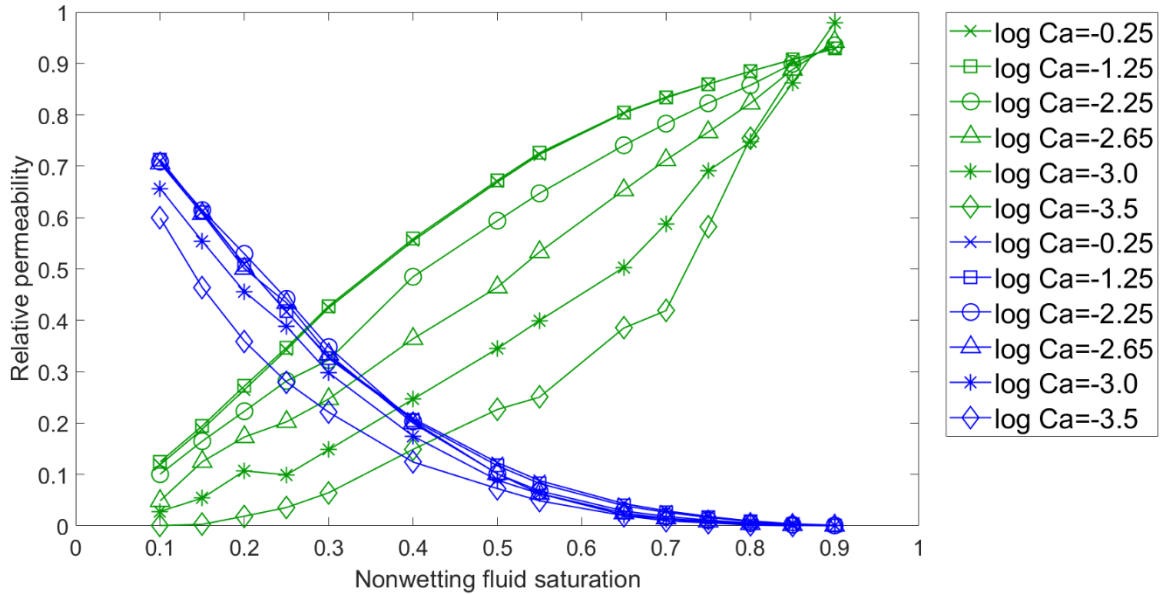


Fig. 7. Relative permeability curves for the nonwetting (k_{nw} , green) and wetting fluids (k_w , blue) for $\log Ca = -0.25$ to -3.5 with $M = 1$.

3.1.2 Effect of Ca change on k_{nw} and k_w

The k_{nw} and k_w curves for six different values of $\log Ca$ (-0.25 , -1.25 , -2.25 , -2.65 , -3.00 , and -3.50 ± 0.10) at $M = 1$ are plotted (Fig. 7). The Ca value is altered by changing the IFT value. It was difficult to keep Ca constant because Ca is also a function of fluid velocity, which is a direct result of the simulation. Thus, we allowed a small error of ± 0.10 for Ca under all conditions.

The k_{nw} values remain relatively constant between $\log Ca = -0.25$ and -1.25 , i.e., for high Ca /low IFT values, and the relative permeability curve for the nonwetting fluid is an approximately straight line, in agreement with previous studies (Asar & Handy, 1989; Shen et

al., 2010). However, as $\log Ca$ becomes lower than -1.25 , k_{nw} notably decreases as Ca decreases. This decrease occurs because the flow of the nonwetting fluid is more strongly inhibited by capillary force when Ca is low. Capillary force is controlled by IFT, the surface tension forces between the nonwetting and wetting fluids. As IFT increases, the nonwetting fluid becomes trapped by the larger capillary force in the small pore spaces, thus causing k_{nw} to decrease. This result also suggests that k_{nw} does not decrease linearly with changes in IFT/Ca , because the k_{nw} decrease gap gradually becomes larger as Ca decreases. This phenomenon occurs because as Ca becomes lower, the influence of capillary force becomes more dominant.

For the wetting fluid, k_w also decreases as Ca decreases (Fig. 7), but at a much smaller rate compared to k_{nw} , consistent with previous research (Harbert, 1983; Fei Jiang et al., 2014; McDougall et al., 2007; Ramstad et al., 2010; Zhao et al., 2017). The reason for this result is that the wetting fluid flows only along the rock surface and the nonwetting fluid flows in the central part of the pores (Fig. 3). Thus, the wetting fluid is mainly affected by interaction with the rock surface, and the change in capillary force has a negligible effect. However, when Ca is low ($\log Ca = -3.50$), the k_w decrease is more notable. This phenomenon can be attributed to the drag force exerted by the nonwetting fluid. At $\log Ca = -3.50$, k_{nw} becomes notably lower compared to other Ca conditions and the nonwetting fluid velocity is low. This result suggests that the influence of the nonwetting fluid drag force on the wetting fluid increases, causing k_w to markedly decrease with the k_{nw} decrease. Nevertheless, we conclude that the variation in k_w is dominated by the fluid saturation in the system and is little affected by changes in IFT or Ca .

3.2 $M-Ca-k_{nw}$ color diagram

After confirming the influence of M and Ca on k_{nw} and k_w in a two-phase flow system, we conducted simulations with various values of M and Ca at a constant 20% S_{nw} to create a M – Ca – k_{nw} correlation diagram (Fig. 8). We chose the 20% S_{nw} condition because it is realistic under all M – Ca conditions (Tsuji et al., 2016). We created the color diagram only for the nonwetting fluid because the relative permeability curves (Fig. 7) demonstrate that k_w does not respond markedly to changes in Ca . We also created a regression model and equation for our results reported in this section (see Appendix B). The maximum Reynolds number reached in this set of simulations was 3.92 for $M = 0.10$ and $\log Ca \approx 0$.

Color maps are useful to provide estimates of k_{nw} under a wide range of M and Ca conditions in a reservoir and to analyze the degree of influence on k_{nw} by M and Ca (Fig. 8). The first color map (Fig. 8a) depicts the parameter space of $\log M = -1.02$ to 0.70 and $\log Ca = -4.00$ to 0.00 , which represents the broad range of conditions that can be simulated. Values higher than this range are hard to achieve at the field scale, and lower values are outside the stable range of the simulation. The second color map (Fig. 8b), covering the lower left quadrant of the parameter range shown in Fig. 8a with parameter space of $\log M = -1.02$ to 0.00 and $\log Ca = -4.00$ to -1.00 , was produced to increase the accuracy of k_{nw} estimates for a system with low M and low Ca , which is common in CCS field.

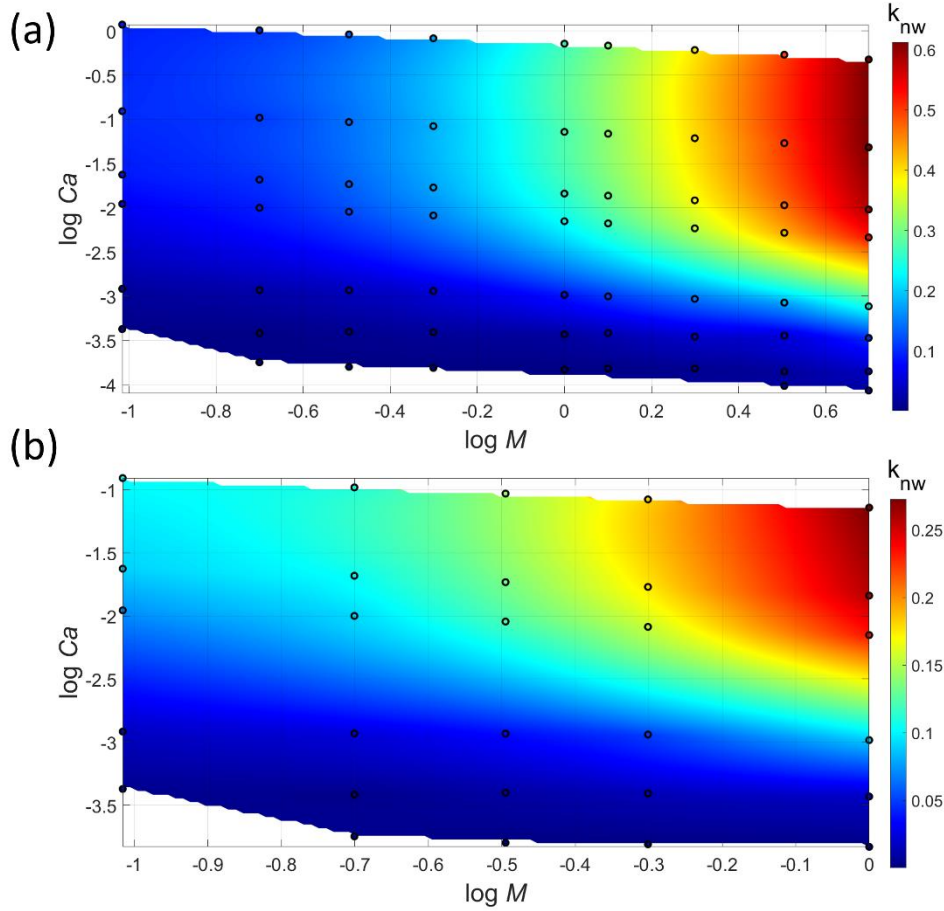


Fig. 8. Relative permeability map of the nonwetting fluid for (a) $\log M = -1.02$ to 0.70 and $\log Ca = -4.00$ to 0.00, and (b) $\log M = -1.02$ to 0.00 and $\log Ca = -4.00$ to -1.00 . The dots indicate the Ca – M conditions used for LBM simulation.

In the specified parameter range, k_{nw} varies from 0.0002 to 0.6125 (Fig. 8a). For high $\log Ca$ values (between -1.25 to 0.00), k_{nw} does not vary with $\log Ca$ and is only influenced by M , as indicated by the almost straight vertical borders between color bands. This finding agrees with our results from section 3.1.2 (Fig. 7), which shows that for $\log Ca \geq -1.25$, the nonwetting fluid relative permeability curve is not affected by Ca . However, as $\log Ca$ becomes lower than -1.25 (Ca value ≈ 0.06), k_{nw} can be seen to vary on both the x axis ($\log M$) and the y axis ($\log Ca$), that is, k_{nw} starts to be affected by the capillary force. The diagram agrees with our finding in section 3.2 that for $\log Ca$ lower than -1.25 , k_{nw} decreases as Ca decreases.

The k_{nw} variation with M (x axis) is more pronounced for $\log M > 0$ ($M > 1$) than for $\log M < 0$ (Fig. 8a). This result is consistent with our findings in section 3.1 and is caused by the lubrication effect resulting from viscosity differences. When μ_{nw} is higher than μ_w , the resulting instability and viscous coupling effect enhance the relative permeability of the nonwetting fluid. The diagram clearly captures the contrast between k_{nw} at $\log M < 0$ and at $\log M > 0$.

To check the detailed k_{nw} variation under low M and low Ca conditions, in Fig. 8b we show an enlargement of part of the lower left quadrant of the parameter range illustrated in Fig. 8a. The k_{nw} value range is 0.0002 to 0.2725 in Fig. 8b, and in this parameter space, both Ca and M influence k_{nw} and thus the borders between the color bands are curved. In a CCS field, M is usually lower than 1 and Ca is generally low (Zheng et al., 2017). Therefore, Fig. 8b is useful for estimating nonwetting fluid relative permeability in a CCS field.

The diagram is useful for evaluating k_{nw} when the effects of M and Ca are considered simultaneously, and the graph neatly summarizes all the main results of this study. The diagram shows that capillary force starts affecting the flow of the nonwetting fluid when $\log Ca$ is less than -1.25 and that the effect of changes in Ca may be just as important as that of changes in M . The graph also shows that k_{nw} is affected by M under all conditions, but most markedly when $\log M > 0$. The graph yields a good estimate of k_{nw} at 20% saturation based on the M and Ca parameters.

4 Discussion

We have demonstrated how M and Ca can markedly change the relative permeabilities of the nonwetting and wetting fluids in a reservoir; thus, our results can help identify the optimum properties of the immiscible fluids to be used in a geologic reservoir. For example, in a CCS

project, the injectivity of the CO₂ is highest when the reservoir fluid has a lower viscosity and the IFT between the two fluids is low. Thus, the relative permeability of injected CO₂ is higher in a saline aquifer than in an oil field. For an EOR project, the relative permeability of oil is higher when a fluid with low viscosity is injected.

The relative permeability map (Fig. 8) is useful to provide accurate estimates of k_{nw} in reservoir-scale simulations. Currently, the relative permeabilities of the nonwetting and wetting fluid are simulated on the basis of a uniform relative permeability curve, without regard to the M and Ca conditions. However, we have shown that the nonwetting fluid, at a typical saturation (S_{nw}) of 20%, can vary in relative permeability by an order of magnitude, from 0.0002 to 0.6125, depending on M and Ca conditions. The color map created in this study can provide more accurate estimates of relative permeability (e.g., temporal permeability variations in a 3D reservoir model) if Ca and M are derived from the reservoir simulation. In addition, although the M generally remains constant in the two-phase flow, the Ca value can greatly changes depending on the distance from the injection well in a reservoir. For example, when evaluating the reservoir area located near the injection well, the injection pressure creates a high-pressure gradient which causes the injected fluid velocity to be high. As the fluid flows away from the injection site, the fluid velocity becomes slower, thus the Ca becomes lower. This means that fluid relative permeability varies based on location inside the reservoir, and the permeability variation can be evaluated using the relative permeability map, thus providing a more accurate relative permeability estimation.

In this study, we conducted the simulation under steady-state conditions, as being most representative for illustrating fluid flows in a reservoir area distant from the injection site, where boundary effects are negligible (Honarpour et al., 2018; Ramstad et al., 2012). In addition, under

steady-state conditions, relative permeability can be measured directly by using Darcy's law; thus, the relative permeability value is more accurate than one calculated under non-steady-state conditions. However, there are some differences between relative permeability values obtained under steady-state and non-steady-state conditions, and there are also some limitations to steady-state simulations. Under steady-state conditions, both fluid phases coexist together in the reservoir at the prescribed saturation from the beginning of the simulation. In contrast, under non-steady-state conditions, the reservoir is initially filled with wetting fluid, and nonwetting fluid is injected and gradually displaces the wetting fluid, which results in a gradual increase of nonwetting fluid saturation and a decrease of wetting fluid saturation. As a result, under non-steady-state conditions several factors can affect relative permeability variation, such as fingering phenomena, drainage and imbibition conditions, and the pore-filling behavior during the injection (Bakhshian et al., 2021; Rabbani et al., 2017). Therefore, an unsteady-state simulation can better represent relative permeability near the borehole or injection site, where remarkable displacement occurs. Nevertheless, if considering the law of two-phase fluid flow that coexist and flow together throughout the whole drainage area, steady-state relative permeability curves are more suitable for planning the whole reservoir design. In addition, relative permeability obtained under steady-state conditions is more accurate because the relative permeability can be directly measured by using Darcy's law. In contrast, in an unsteady-state simulation, relative permeability must be calculated from the capillary pressure change and production data under several assumptions (e.g., the Johnson, Bossler, and Naumann method; (Esmaeili et al., 2020; Johnson et al., 1959)). As a result, the obtained relative permeability is less accurate. Steady-state condition is also better to estimate relative permeability for CCS

monitoring system after injection, as at the time, the CO₂ will coexist with the wetting fluid in the reservoir.

In this study, we created a relative permeability map for $S_{nw} = 20\%$, which is a reasonable condition that can be achieved in all or most systems. As S_{nw} increases, k_{nw} also increases until it reaches its maximum value (e.g., k_{nw} at the irreducible saturation). Therefore, it is advisable to create relative permeability maps for several saturation conditions. One possible future direction from this study would be to create a four-dimensional M – Ca – S_{nw} – k_{nw} graph to yield k_{nw} estimates for all saturation conditions. Nevertheless, the great consistency of variations in k caused by changes in M and Ca found in this study suggests that maps for other saturation conditions will have similar features to Fig. 8. It would also be possible to create a regression model for $k_{nw} = f(M-Ca-S_{nw})$ to estimate k_{nw} under all saturation conditions.

In this study, we produced a relative permeability diagram for a digital specimen of Berea sandstone. In addition to M and Ca , the influences upon relative permeability include the pore geometry of the rock, such as pore size distribution, pore connectivity, and other parameters (Jiang et al. 2018, WRR). These parameters differ among rock formations, meaning that the relative permeability maps of different types of reservoir rocks will vary. Our methodology makes it possible to create accurate maps of relative permeability for other reservoir rocks.

In this study, we used $\log M$ ranging from -1.02 to 0.70 and $\log Ca$ ranging from -4.00 to 0.00 ; however, in the CO₂ injection process, $\log M$ might reach as low as -1.8 and $\log Ca$ might reach -6.8 (Zheng et al., 2017). Such values are outside the stable range where this simulation can be applied. Nevertheless, the objective of this study was just to propose a method to predict the relative permeability of a two-phase flow fluid in a reservoir by using a color map diagram to evaluate the effect of the viscosity ratio and capillary number (Fig. 8). In a future

study, we plan to improve the stability range of the simulation to allow lower M and Ca values to be considered.

In geological CO_2 storage, the maximum amount of CO_2 that can be stored and the injectivity of the CO_2 into the reservoir must both be considered (Tsuji et al., 2016). Thus, the results of this study must be combined with information on the effects of M and Ca on the maximum saturation of the nonwetting fluid. In this study, we showed that, for CO_2 as the nonwetting fluid, the relative permeability increases as M increases and decreases as Ca becomes very small. In contrast, Tsuji et al. (2016) demonstrated that the maximum S_{nw} increases as M increases and notably increases at low Ca . Thus, although a high value of M is desirable to increase both the CO_2 capacity and injectivity, a low Ca value can also increase the maximum CO_2 saturation but at the cost of reduced relative permeability. Both factors must be taken into account when choosing suitable conditions for CO_2 storage.

An advantage of using M and Ca is that both parameters are dimensionless, meaning that the results obtained in this pore-scale study can potentially be upscaled to the reservoir scale. Ideally, the pore-scale results are also valid at the reservoir scale as long as the ratios of the parameters (viscosity and IFT) are maintained for both fluids, because the fluid flow behavior at the reservoir scale is controlled by the fluid dynamics at the pore scale. However, this ideal is challenged by the inhomogeneity of the porous medium. The relative permeability is likely to vary throughout the reservoir due to differences in pore size, pore connectivity, and many other factors. Nevertheless, the results of a pore-scale simulation are important to verify relative permeability variations arising from selected factors (in this study, M and Ca) by eliminating other factors. The results of a pore-scale study of relative permeability could then be upscaled by

considering the structural factors of the reservoir, e.g., its porosity and pore connectivity, using advanced techniques such as machine learning.

5 Summary

To evaluate the influence of viscosity ratio M and capillary number Ca on relative permeability k in a two-phase flow system, we calculated k for nonwetting and wetting fluids (k_{nw} and k_w) under various M and Ca conditions using an LBM simulation. The main results of this study are as follows.

- 1) In our simulations, the relative permeability of the nonwetting fluid increased as the viscosity ratio increased due to the lubricating effect, locally occurred pore-filling behavior, and instability of the fluid interface. Specifically, at high viscosity ratios ($M = 5$), k_{nw} could exceed 1 as a result of the lubricating effect.
- 2) The relative permeability of the wetting fluid decreased as M increased due to the increase in shear force from the nonwetting fluid (viscous coupling effect)
- 3) As M increased, the number nonwetting fluid clusters became higher, indicating that the nonwetting fluid became more disconnected and the size of each clusters becomes smaller.
- 4) At high capillary numbers ($\log Ca = -0.25$ to -1.25), k_{nw} did not respond to Ca , however as $\log Ca$ becomes lower than -1.25 , k_{nw} decreases as Ca decreases due to the capillary force effect.
- 5) As Ca decreased k_w decreased, but at a negligible rate compared to k_{nw} .

6) k_{nw} can change markedly in a wide range of M – Ca parameter space, and the M – Ca – k_{nw} correlation map created in this study can provide k_{nw} estimates at various reservoir conditions.

Acknowledgements

This study was supported by the Japan Society for the Promotion of Science (JSPS) through a Grant-in-Aid for Challenging Exploratory Research (Grant Number JP20K20948). This work was also partially supported by the JSPS KAKENHI Grant Number JP19K15100. We are grateful for the funding provided by the Top Global University project conducted by the Ministry of Education, Culture, Sports, and Technology, Japan (MEXT). We also gratefully acknowledge support of International Institute for Carbon-Neutral Energy Research (I²CNER). The Micro-CT data used to reconstruct the digital rock model is achieved by The Imperial College Consortium on Pore-Scale Modelling and Imaging.

Appendices

A. Single-phase simulation

Before conducting the two-phase flow simulations, we first ran single-phase simulations to calculate the absolute permeability of the 3D rock model. Single-phase simulations can also be used to verify the accuracy and precision of the simulation. When only one type of fluid exists in a rock, the absolute permeability can be calculated from Darcy's law:

$$k = \frac{v \mu \Delta l}{\Delta P} \quad (\text{A. 1})$$

where k is the absolute permeability (m^2), v is the average fluid velocity (m/s), μ is the dynamic viscosity of the fluid (Pa s), Δl is the distance between the inlet and the outlet, and ΔP is the

applied pressure difference between the inlet and the outlet. In this study, because the body force was applied only in the z direction, we calculated k in the z direction (k_z).

Because all simulation units are in lattice Boltzmann units, they must subsequently be converted into physical units. The unit conversion for k can be performed using the formula:

$$k_{(physical)} = k_{(LB)} * A^2 \quad (A. 2)$$

where A is the resolution of the digital rock, i.e. the physical size of a lattice grid (2.673 μm in this study).

Absolute permeability is an intrinsic property of a rock that is independent of the type of fluid. The single-phase simulations were conducted with several different values of fluid viscosity and body force to ensure the consistency of the results. When the fluid summation of velocity change in 1,000 steps difference was less than 2% for all cases, the simulations were assumed to have converged at that time. The absolute permeability results in various viscosity and body force conditions are shown in the table A.1.

kinematic viscosity (lattice-Boltzmann unit)	Body force (lattice-Boltzmann unit)	absolute permeability (Darcy)
0.155555	0.0001	1.3029
0.155555	0.0002	1.3025
0.310000	0.0001	1.3042
0.310000	0.0002	1.3038

Table A.1 single-phase simulation conditions and results

The absolute permeability of the Berea rock is 1.3034 ± 0.001 Darcy. The absolute permeability value remains constant under various conditions, which demonstrates the accuracy and stability of the simulation. The result obtained from this simulation is slightly different from

the result by the Dong and Blunt (2009), which gives the absolute permeability value of 1.193 Darcy. The difference might be due to the difference in size of digital rocks. Dong and Blunt used digital rock with dimension of 400x400x400 voxels with resolution of 5.345 μm (9.772 mm^3 in actual size), while we used digital rock with dimension of 400x400x400 voxels with resolution of 2.673 μm (1.222 mm^3 in actual size) is used. The dimension difference might have caused a slight change in the rock sample heterogeneity. In addition, Dong and Blunt applied Pore Network Model (PNM) to calculate the absolute permeability, while in this study, LBM simulation is used. Nonetheless, the absolute permeability value obtained is still in the same order as described by the paper.

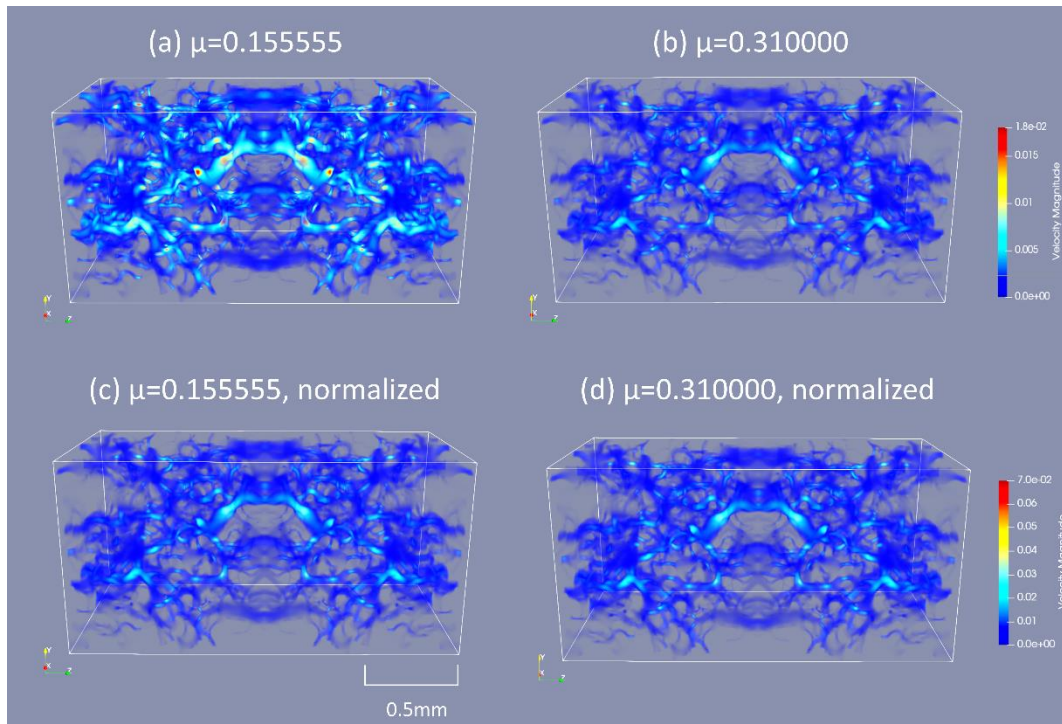


Fig. A. Velocity fields for single phase simulation at (a) $\mu = 0.155555$ and (b) $\mu = 0.310000$ and normalized velocity fields for single phase simulation at (c) $\mu = 0.155555$ and (d) $\mu = 0.310000$.

We show the velocity field of single-phase simulation for kinematic viscosity = 0.155555 and 0.310000 with equal body force of 0.0001 (Fig. A(a) and A(b)). The fluid velocity for simulation with fluid kinematic viscosity of 0.155555 nu_{LB} is significantly higher compared to the fluid kinematic viscosity of 0.310000 nu_{LB} case. This is because fluid average velocity is inversely proportional to viscosity. Thus, fluid velocity is not proportional to permeability, and must be normalized into dimensionless velocity field. Fig. A(c) and A(d) show the normalized fluid velocity field for both conditions. The normalization is done using the equation $U^* = \frac{U}{\Delta PL/\mu}$. Both figures show a relatively similar results, which indicate that a similar absolute permeability value is obtained from both conditions.

B. Regression model of the M - Ca - k_{nw} relationship

We created a regression model to generate an empirical equation that can be used to calculate k_{nw} as a function of M and $\log Ca$ at 20% S_{nw} . A height map of our simulation results and the fitted polynomial regression curve is shown in Fig. B.

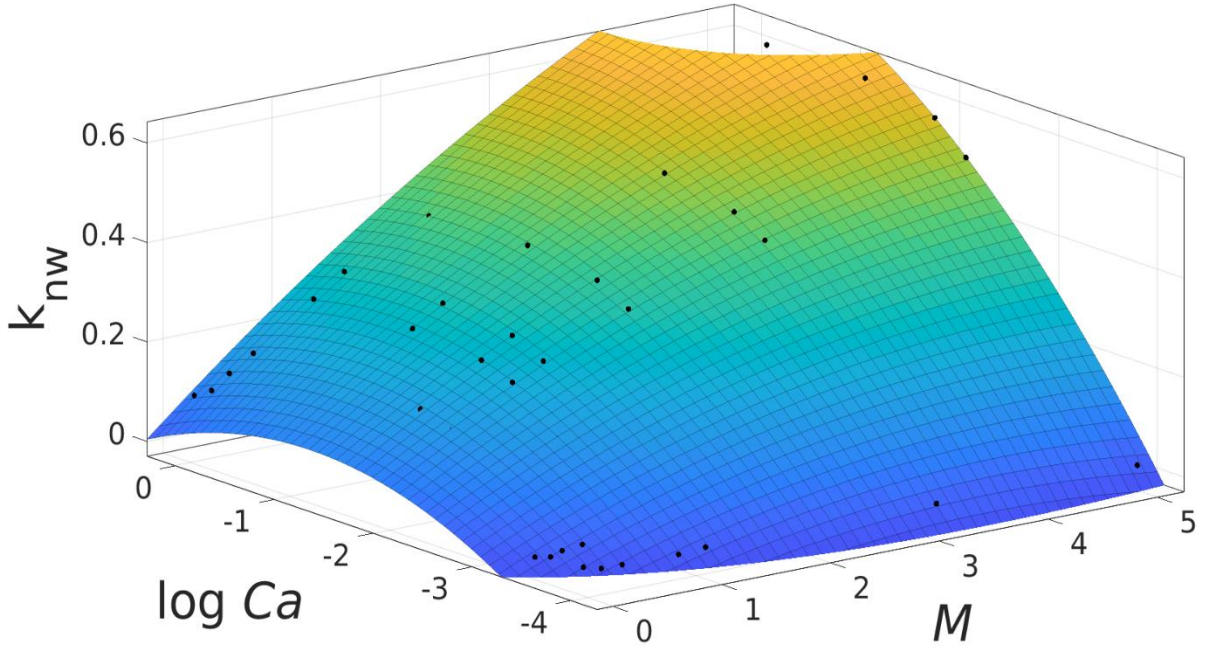


Fig. B. Nonwetting fluid relative permeability height map (dots) and the fitted polynomial regression curve.

Past research has shown that k_{nw} can be estimated as a function of M by using an equation with second-degree polynomial (equation 3 and 4 (Goldsmith & Mason, 1963)); thus, we chose a second polynomial equation for the regression towards the M value. For the $\log Ca$ value, we also set it as second-degree polynomial to prevent overfitting.

Based on our model, k_{nw} can be calculated with the following formula:

$$k_{nw} = 0.05983 + 0.1844 M - 0.08714 (\log Ca) - 0.009922 M^2 + 0.0275M(\log Ca) - 0.03335 (\log Ca)^2$$

Our model has an R -squared value of 0.9481 and a summed square of residuals value of 0.09871.

Because we created a color map diagram for only one saturation condition ($S_{nw} = 20\%$), we only created a regression model to calculate k_{nw} based on M and Ca values for this saturation condition. Because the relative permeability increase with respect to saturation is not linear, we expect the regression model to be different for other saturation conditions. In the future, we plan to create more relative permeability color maps for various saturation conditions and create a four-dimensional $\log M$ – $\log Ca$ – S_{nw} – k_{nw} graph to yield k_{nw} estimates for all saturation conditions. Then, we will attempt to create a regression model to estimate k_{nw} as a function of M , $\log Ca$, and S_{nw} .

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Figure Caption

Fig. 1. Berea sandstone digital rock model: (a) core form and (b) mirrored form.

Fig. 2. Relative permeability curves as a function of M at constant $\log Ca = -0.25 \pm 0.1$ for $M = 0.1-5$: (a) nonwetting fluid relative permeability (k_{nw}) curve and (b) wetting fluid relative permeability (k_w) curve.

Fig. 3. Two-dimensional slice showing the distributions of nonwetting fluid (red), wetting fluid (green), and solid rock (black) at various values of M and S_{nw} .

Fig. 4. Evolution of the number of clusters of nonwetting fluid at $M = 1$ and $M = 5$, with $S_{nw} = 20\%$ and $\log Ca = -0.25 \pm 0.1$.

Fig. 5. Two-dimensional slice showing the distributions of nonwetting (red), wetting fluid (green), and solid rock (black) at (a) $M = 1$ and (b) $M = 5$ at 20% S_{nw} . These images show a portion of the slice illustrated in Fig. 3.

Fig. 6. Normalized velocity fields for the nonwetting fluid in a two-phase simulation with (a) $M = 1$ and (b) $M = 5$, at $\log Ca = -0.25 \pm 0.1$ and $S_{nw} = 20\%$.

Fig. 7. Relative permeability curves for the nonwetting (k_{nw} , green) and wetting fluids (k_w , blue) for $\log Ca = -0.25$ to -3.5 with $M = 1$.

Fig. 8. Relative permeability map of the nonwetting fluid for (a) $\log M = -1.02$ to 0.70 and $\log Ca = -4.00$ to 0.00 , and (b) $\log M = -1.02$ to 0.00 and $\log Ca = -4.00$ to -1.00 . The dots indicate the $Ca-M$ conditions used for LBM simulation.

Fig. A. Velocity fields for single phase simulation at (a) $\mu = 0.155555$ and (b) $\mu = 0.310000$ and normalized velocity fields for single phase simulation at (c) $\mu = 0.155555$ and (d) $\mu = 0.310000$.

Fig. B. Nonwetting fluid relative permeability height map (dots) and the fitted polynomial regression curve.