# Adaptive Time-Splitting Scheme for Nanoparticles Transport with Two-phase Flow in Heterogeneous Porous Media

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Abstract. In this work, we introduce an efficient scheme using an adaptive time-splitting method to simulate nanoparticles transport associated with a two-phase flow in heterogeneous porous media. The capillary pressure is linearized in terms of saturation to couple the pressure and saturation equations. The governing equations are solved using an IMplicit Pressure Explicit Saturation-IMplicit Concentration (IMPES-IMC) scheme. The spatial discretization has been done using the cellcentered finite difference (CCFD) method. The interval of time has been divided into three levels, the pressure level, the saturation level, and the concentrations level, which can reduce the computational cost. The time step-sizes at different levels are adaptive iteratively by satisfying the Courant-Friedrichs-Lewy (CFL<1) condition. The results illustrates the efficiency of the numerical scheme. A numerical example of a highly heterogeneous porous medium has been introduced. Moreover, the adaptive time step-sizes are shown in graphs.

Keywords: Time-splitting  $\cdot$  IMPES  $\cdot$  Two-phase flow  $\cdot$  Porous media  $\cdot$  CFL.

## 1 Introduction

The scheme of IMplicit Pressure Explicit Saturation (IMPES) is a conditionally stable which is usually used to solve the two-phase flow in porous media. In the IMPES scheme, the pressure equation is treated implicitly while the saturation is updated explicitly. Hence it takes a very small time step size, in particular with heterogeneous porous media. The IMPES scheme has been improved in several versions (e.g. [1–3]). The temporal discretization scheme is considered an important factor that affects the efficiency of numerical reservoir simulators. The application of traditional single-scale temporal scheme is restricted by the rapid changes of the pressure and saturation with capillarity and concentrations

if applicable. Applying time splitting strategies has a significant improvement compare to the traditional schemes. Time splitting method has been considered in a number of publications such as [4]-[10]. For example, in Refs. [9], an explicit sub-timing scheme is provided. On the other hand, an implicit time-stepping scheme have been proposed by Bhallamudi et al. [7] and Park et al. [8].

In the recent years, nanoparticles have been used in many engineering branches including petroleum applications such as enhanced oil recovery. A model of nanoparticles transport in porous media have been established by Ju et al. [13] based on particle migration in porous media model [14]. El-Amin et al. [15–17] have investigated the problem of nanoparticles transport in porous media. In Ref. [17], they presented an extended model to include mixed relative permeabilities and a negative capillary pressure. Dimensional analysis of the problem of nanoparticles transport in porous media have been presented by El-Amin et al. [18]. The nanoparticles transport in anisotropic porous media have been studied numerically using the multipoint flux approximation by Salama et al. [19]. Also, a numerical simulation of drag reduction effects by hydrophobic nanoparticles adsorption method in water flooding processes has been presented by Chen et al. [20]. The problem of dynamic update of an anisotropic permeability field with nanoparticles transport in porous media has been considering by Chen et al. [21]. In Ref. [22], the authors presented a nonlinear iterative IMPES-IMC scheme to solve the governing system of the nanoparticles transport in subsurface.

In this paper, we propose a time-stepping IMPES-IMC scheme to solve the governing equations of the nanoparticles transport associated with two-phase flow in subsurface. The CCFD method has employed for the spatial discretization. The time-splitting scheme has applied together with the CFL condition to adaptive the time-steps sizes. Finally, some numerical experiments are provided.

# 2 Modeling and Mathematical Formulation

The mathematical model of the problem under consideration consists of water saturation, Darcy's law, nanoparticles concentration, deposited nanoparticles concentration on the pore-wall, and entrapped nanoparticles concentration in the pore-throat. Moreover, the variations in both porosity and permeability due to the nanoparticles deposition/entrapment on/in the pores have been taken into consideration. In the following we introduce the governing equations briefly, (for details see Refs. [15–19, 21, 22]:

Momentum Conservation (Darcy's Law):

$$\mathbf{u}_{\alpha} = -\frac{k_{\alpha}}{\mu_{\alpha}} \nabla \Phi_{\alpha}, \qquad \alpha = w, n, \tag{1}$$

where

$$k_{\alpha} = k_{r\alpha} \mathbf{K}, \quad \Phi_{\alpha} = p_{\alpha} + \rho_{\alpha} g \nabla z, \quad \alpha = w, n,$$

#### Mass Conservation (Saturation Equations):

$$\phi \frac{\partial s_{\alpha}}{\partial t} + \nabla \cdot \mathbf{u}_{\alpha} = q_{\alpha}, \quad \alpha = w, n, \tag{2}$$

where  $s_w + s_n = 1$ ,  $\phi$  is the porosity, g is the gravitational acceleration, z is the depth.  $\mathbf{u}_{\alpha}$  is the velocity,  $\Phi_{\alpha}$  is the the pressure potential,  $p_{\alpha}$  is the pressure,  $\mu_{\alpha}$  is the viscosity,  $\rho_{\alpha}$  is the density,  $k_{r\alpha}$  is the relative permeability,  $q_{\alpha}$  is the external mass flow rate,  $s_{\alpha}$  is the saturation; all of the phase  $\alpha$ . **K** is the permeability tensor. w stands for the wetting phase (water), and n stands for the non-wetting phase (oil).

Providing the following definitions:

The capillary pressure:  $p_c(s_w) = p_n - p_w$ .

The total velocity:  $\mathbf{u}_t = \mathbf{u}_w + \mathbf{u}_n$ .

The flow fraction:  $f_w = \lambda_w / \lambda_t$ .

The phase mobility:  $\lambda_{\alpha} = k_{r\alpha}/\mu_{\alpha}$ .

The total mobility:  $\lambda_t$ .

The capillary pressure potential:  $\Phi_c = p_c + (\rho_n - \rho_w) g \nabla z$ .

The total source mass transfer:  $q_t = q_w + q_n$ .

After some mathematical manipulations and referring to Refs. [11], the pressure equation can be rewritten as,

$$\nabla \cdot \mathbf{u}_t = -\nabla \cdot \lambda_t \mathbf{K} \nabla \Phi_w - \nabla \cdot \lambda_n \mathbf{K} \nabla \Phi_c = q_t.$$
(3)

Because this equation contents the capillary pressure which is a function of saturation, it will be coupled with the following saturation equation to calculate the pressure,

$$\phi \frac{\partial s_w}{\partial t} - q_w = -\nabla \cdot \lambda_w \mathbf{K} \nabla \Phi_w. \tag{4}$$

However, the saturation is updated using the following form,

$$\phi \frac{\partial s_w}{\partial t} - q_w = -\nabla \cdot (f_w \mathbf{u}_a). \tag{5}$$

where  $\mathbf{u}_w = f_w \mathbf{u}_a$  and  $\mathbf{u}_a = -\lambda_t \mathbf{K} \nabla \Phi_w$ .

#### Nanoparticles Transport Model:

Assuming that the nanoparticles exist only in the water phase of one size interval. So, The transport equation of the nanoparticles in the water phase is given as ([12, 14, 13, 15–19, 21, 22]),

$$\phi \frac{\partial \left(s_w c\right)}{\partial t} + \nabla \cdot \left(\mathbf{u}_w c - D\nabla c\right) = R + Q_c,\tag{6}$$

### Nanoparticles Surface Deposition:

The surface deposition is expressed by,

$$\frac{\partial c_{s1}}{\partial t} = \begin{cases} \gamma_d |\mathbf{u}_w|c, \quad \mathbf{u}_w \le u_r \\ \gamma_d |\mathbf{u}_w|c - \gamma_e |\mathbf{u}_w - u_r|c_{s1}, \, \mathbf{u}_w > u_r \end{cases}$$
(7)

#### Nanoparticles Throat Entrapment:

The rate of entrapment of the nanoparticles is,

$$\frac{\partial c_{s2}}{\partial t} = \gamma_{pt} |\mathbf{u}_w| c, \tag{8}$$

where c is the nanoparticles concentrations.  $c_{s1}$  and  $c_{s2}$  are, respectively, the deposited nanoparticles concentration on the pore surface, and the entrapped nanoparticles concentration in pore throats.  $\tau$  is the tortuosity parameter.  $Q_c$  is the rate of change of particle volume belonging to a source/sink term.  $\gamma_d$  is the rate coefficients for surface retention of the nanoparticles.  $\gamma_e$  is the rate coefficients for entrainment of the nanoparticles.  $u_r$  is the critical velocity of the water phase. where  $\gamma_{pt}$  is the pore throat blocking constants. The diffusion-dispersion tensor is defined by,

$$D = \phi s_w \tau D_t, \quad D_t = D^{\mathrm{Br}} + D^{\mathrm{disp}} \tag{9}$$

where  $D^{\text{Br}}$  is the Brownian diffusion and  $D^{\text{disp}}$  is the dispersion coefficient which is defined by [1],

$$\phi s_w \tau D^{\text{disp}} = d_{t,w} |\mathbf{u}_w| \mathbf{I} + (d_{l,w} - d_{t,w}) \frac{\mathbf{u}_w \mathbf{u}_w^T}{|\mathbf{u}_w|}$$
(10)

Thus,

$$D = (\phi s_w \tau D^{\mathrm{Br}} + d_{t,w} |\mathbf{u}_w|) \mathbf{I} + (d_{l,w} - d_{t,w}) \frac{\mathbf{u}_w \mathbf{u}_w^{\mathrm{I}}}{|\mathbf{u}_w|}$$
(11)

where  $d_{l,w}$  and  $d_{t,w}$  are the longitudinal and transverse dispersion coefficients, respectively. R is the net rate of loss of nanoparticles which is defined by,

$$R = \frac{\partial c_{s1}}{\partial t} + \frac{\partial c_{s2}}{\partial t} \tag{12}$$

 $\mathbf{T}$ 

#### Initial and Boundary Conditions:

The initial conditions are,

$$s_w = s_w^0, \quad c = c_{s1} = c_{s2} = 0 \quad \text{in} \quad \Omega \qquad \text{at} \quad t = 0,$$
 (13)

The boundary conditions are given as,

$$p_w (\text{or } p_n) = p^D \quad \text{on} \quad \Gamma_D,$$
 (14)

$$\mathbf{u}_t \cdot \mathbf{n} = q^N, \quad s_w = S^N, \quad c = c_0, \quad c_{s1} = c_{s2} = 0 \quad \text{on} \quad \Gamma_N.$$
 (15)

where **n** is the outward unit normal vector to  $\partial\Omega$ ,  $p^D$  is the pressure on  $\Gamma_D$  and  $q^N$  the imposed inflow rate on  $\Gamma_N$ , respectively.  $\Omega$  is the computational domain such that the boundary  $\partial\Omega$  is Dirichlet  $\Gamma_D$  and/or Neumann  $\Gamma_N$  boundaries, i.e.  $\partial\Omega = \Gamma_D \cup \Gamma_N$  and  $\Gamma_D \cap \Gamma_N = \emptyset$ .

# 3 Multi-scale Time Splitting Method

The concept of time splitting method is to use different time step size for each equation has a time derivative. In the above-described method, the pressure is

coupled with the saturation in each time-step. The time-step size for the pressure can be taken larger than the those of saturation and nanoparticle concentrations. So, for the pressure the total time interval [0,T] is divided into  $N_p$ , time-steps as  $0 = t_0 < t_1 < \cdots < t_{N_p=T}$ . Thus, the time-step length assigned for pressure is,  $\Delta t^k = t^{k+1} - t^k$ . Since the saturation varies more rapidly than the pressure, we use a smaller time-step size for the saturation equation. That is, each interval,  $(t^k, t^{k+1}]$ , will be divided into  $N_{p,s}$  subintervals as  $(t^k, t^{k+1}] = \bigcup_{l=0}^{N_{p,s}-1}(t^{k,l}, t^{k,l+1}]$ . On the other hand, as the concentration varies more rapidly than the pressure (and may be saturation), we also use a smaller time-step size for the concentration equations. Thus, we partition each subinterval  $(t^{k,l}, t^{k,l+1}]$  into  $N_{p,s,c}$  subsubintervals as  $(t^{k,l}, t^{k,l+1}] = \bigcup_{m=0}^{N_{p,s,c}-1}(t^{k,l,m+1}]$ . Therefore, the system of governing equations, (3), (4), (6), (7) and (8), is solved based on the adaptive time-splitting technique. The backward Euler time discretization is used for the equations of pressure, saturation, concentration and the two volume concentration. We linearized the capillary pressure function,  $\Phi_c$ , in terms of saturation using this formula,

$$\Phi_c\left(s_w^*\right) \cong \Phi_c\left(s_w^k\right) + \Phi_c'\left(s_w^k\right) \left[s_w^{k+1} - s_w^k\right],\tag{16}$$

where  $\Phi'_c$  is derivative of  $\Phi_c$ . The quantity,  $[s_w^{k+1} - s_w^k]$ , can be calculated from the saturation equation,

$$s_{w}^{k+1} - s_{w}^{k} = \frac{\Delta t^{k}}{\phi\left(c_{s_{1}}^{k}, c_{s_{2}}^{k}\right)} \left[q_{w}^{k+1} - \nabla \cdot \lambda_{t}\left(s_{w}^{k}\right) \mathbf{K}\left(c_{s_{1}}^{k}, c_{s_{2}}^{k}\right) \nabla \Phi_{w}^{k+1}\right].$$
(17)

In addition to the pressure equation,

$$-\nabla \cdot \lambda_t \left(s_w^k\right) \mathbf{K} \left(c_{s_1}^k, c_{s_2}^k\right) \nabla \Phi_w^{k+1} - \nabla \cdot \lambda_n \left(s_w^k\right) \mathbf{K} \left(c_{s_1}^k, c_{s_2}^k\right) \nabla \Phi_c \left(s_w^*\right) = q_t^{k+1} (18)$$

Then, the above coupled system (16), (17) and (18) is solved implicitly to obtain the pressure potential. Therefore, the saturation is updated explicitly with using the upwind scheme for the convection term as,

$$\phi\left(c_{s1}^{k}, c_{s2}^{k}\right) \frac{s_{w}^{k,l+1} - s_{w}^{k,l}}{\Delta t^{l}} + \nabla \cdot \left(f_{w}^{k} \mathbf{u}_{a}^{k+1}\right) = q_{w}^{k,l+1}.$$
(19)

Therefore, the nanoparticles concentration, deposited nanoparticles concentration on the pore–walls and entrapped nanoparticles concentration in the pore– throats are computed implicitly as follow,

$$\phi\left(c_{s1}^{k}, c_{s2}^{k}\right) \frac{s_{w}^{k+1} c^{k,l,m+1} - s_{w}^{k} c^{k,l,m}}{\Delta t^{m}} + \nabla \cdot \left\{\mathbf{u}_{w}^{k+1} c^{k,l,m+1} - D\left(s_{w}^{k+1}, \mathbf{u}_{w}^{k+1}, c_{s1}^{k}, c_{s2}^{k}\right) - R\left(\mathbf{u}_{w}^{k+1}, c_{s1}^{k}\right)\right\} \nabla c^{k,l,m+1} = Q_{c}^{k,l,m+1},$$

$$(20)$$

$$\frac{c_{s_1}^{k,l,m+1} - c_{s_1}^{k,l,m}}{\Delta t^m} = \begin{cases} \gamma_d |\mathbf{u}_w^{k+1}| c^{k+1}, & \mathbf{u}_w^{k+1} \le u_r \\ \gamma_d |\mathbf{u}_w^{k+1}| c^{k+1} - \gamma_e |\mathbf{u}_w^{k+1} - u_r| c_{s_1}^{k,l,m+1}, \mathbf{u}_w^{k,l,m+1} > u_r \end{cases}$$
(21)

and,

$$\frac{c_{s2}^{k,l,m+1} - c_{s2}^{k,l,m}}{\Delta t^m} = \gamma_{pt} |\mathbf{u}_w^{k+1}| c^{k,l,m+1}$$
(22)

Finally, other parameters such as permeability, porosity,  $\lambda_w, \lambda_n, \lambda_t$  and  $f_w$  are updated each loop.

## 4 Adaptive Time-Stepping

In our algorithm, we have checked the Courant–Friedrichs–Lewy (CFL) condition to guarantee its satisfactory (i.e. CFL<1). In order to achieve this idea, we have to define the following CFLs,

$$\operatorname{CFL}_{s,x} = \frac{\mathbf{u}_x \Delta t^{k,l}}{\Delta x}, \quad \operatorname{CFL}_{s,y} = \frac{\mathbf{u}_y \Delta t^{k,l}}{\Delta y},$$
 (23)

for saturation equation, and,

$$\operatorname{CFL}_{c,x} = \frac{\mathbf{u}_x \Delta t^{k,l,m}}{\Delta x}, \quad \operatorname{CFL}_{c,y} = \frac{\mathbf{u}_y \Delta t^{k,l,m}}{\Delta y},$$
 (24)

and for concentration equations. It may be noted that the CFL depends on the ratio  $\Delta t/\Delta x$  which can be fixed at larger time-steps and larger mesh-size. Thus, when we use a lager domain (larger mesh size) then we can use larger time step size. In the code, the initial time-step for the saturation equation is taken as the pressure time-step, i.e.,  $\Delta t^{k,0} = \Delta t^k$ , and the initial time-step for the concentration equation is taken as the saturation time-step, i.e.,  $\Delta t^{k,l,0} = \Delta t^{k,l}$ . Then, we check if  $\text{CFL}_{s,x} > 1$  or  $\text{CFL}_{s,y} > 1$ , the saturation time-step will be divided by 2 and the  $\text{CFL}_{s,x}$  and  $\text{CFL}_{s,y}$  will be recalculated. This procedure will be repeated until satisfying the condition  $\text{CFL}_{s,x} < 1$  and  $\text{CFL}_{s,y} < 1$ , then the final adaptive saturation time-step will be obtained. Similarly, we check if the condition,  $\text{CFL}_{c,x} > 1$  or  $\text{CFL}_{c,y} > 1$  is satisfied, the concentration time-step will be divided by 2 and therefore, we recalculate both  $\text{CFL}_{c,x}$  and  $\text{CFL}_{c,y}$ . We repeat this procedure to reach the condition  $\text{CFL}_{c,x} < 1$  and  $\text{CFL}_{c,y} < 1$ , then we obtain the final adaptive concentration time-step.

## 5 Numerical Tests

In order to examine the performance of the current scheme, we introduce some numerical examples in this section. Firstly, we introduce the required physical parameters used in the computations. Then, we study the performance of the scheme by introducing some results for the adaptive time steps based on values of the corresponding Courant–Friedrichs–Lewy (CFL). Then we present some results for the distributions of water saturation and nanoparticles concentrations. In this study, given the normalized wetting phase saturation,  $S = (s_w - s_{wr})/(1 - s_{nr} - s_{wr}), 0 \le S \le 1$ , the capillary pressure is defined as,  $p_c = -p_e \log S$ , and

the relative permeabilities are defined as,  $k_{rw} = k_{rw}^0 S^2$ ,  $k_{rn} = k_{rn}^0 (1-S)^2$ ,  $k_{rw}^0 = k_{rw} (S=1)$ ,  $k_{rw}^0 = k_{rw} (S=1)$ .  $p_e$  is the capillary pressure parameter,  $s_{wr}$  is the irreducible water saturation and  $s_{nr}$  is the residual oil saturation after water flooding. The values and units of the physical parameters are inserted in Table 1.

In Table 2, some error estimates for water saturation, nanoparticles concentration, and deposited nanoparticles concentration, are provided for various values of the number of time steps k. The reference case  $(s_{w,r}^{n+1} = 0.9565, c_r^{n+1} = 0.0360, \text{ and } c_{s1,r}^{n+1} = 1.3233 \times 10^{-7})$  is calculated at the point (0.06,0.053) for non-adaptive dense mesh of  $200 \times 150$  for  $0.6 \times 0.4$  m, at 2000 time step. It can be seen from Table 2 that the error decreases as number of number of time steps increases. Table 3 shows the real cost time of running the adaptive scheme for different values of the number of time steps k. One may notice from this table that the real cost time decreases as the number of time steps k increases. It seems more time is needed for the adaptation process.

Table 1. Values of the physical parameters.

Parameter	Value	Units
$\gamma_d, \gamma_{pt}, \gamma_e$	16, 1.28, 30	$m^{-1}$
$\mathbf{u}_r$	$4.6 \times 10^{-6}$	m/s
$c_0$	0.1	_
$S_{wr}, S_{nr}$	0.001	_
$\phi_0$	0.3	_
$k_f$	0.6	_
$\gamma_f$	0.01	_
$\mu_w, \mu_n$	1,0.45	$cP = 1.0 \times 10^{-3} Pa.s$
$k_{rw0}, k_{ro0}$	1	_
$B_c$	50	bar = $1.0 \times 10^5$ Pa
D	$4.6 \times 10^{-8}$	$m^2/s$

Table 2. Error estimates for various values of number of time steps k.

k	$\ s_w^{n+1,k+1} - s_{w,r}^{n+1}\ $	$\ c^{n+1,k+1} - c^{n+1}_r\ $	$\ c_{s1}^{n+1,k+1} - c_{s1,r}^{n+1}\ $
500	0.0149	0.0065	1.8701E-05
200	0.0151	0.0065	1.8366E-05
100	0.0166	0.0068	9.9085 E-05

We use a real permeability map of dimensions  $120 \times 50$ , which is extracted from Ref. [23]. The permeabilities vary in a large scope and are they highly heterogeneous. We consider a domain of size  $40m \times 16m$  which is discretized into  $120 \times 50$  uniform rectangles grids. The injection rate was 0.01 Pore-Volume-Injection (PVI) and continued the calculation until 0.5 PV. In this example, we

**Table 3.** Real cost time of running the adaptive scheme for different values of the number of time steps k.

k	100	300	500	750
Real time [s]	3215	2307	532	424

choose the number of steps of the outer loop to be: (Case 1, k=50, Fig. 1; Case 1, k=100, Fig. 1; Case 1, k=200, Fig. 3). In these figures, the adaptive time-step sizes,  $\Delta t^l$  and  $\Delta t^m$  are plotted against the number of steps of the outer loop k and the number of the inner loops l and m. It can be seen from these figures (Fig. 1 – Fig. 3) that the behavior of adaptive  $\Delta t^l$  and  $\Delta t^m$  are very similar. This may be because the velocity is large and dominate the CFL. Also, both  $\Delta t^l$  and  $\Delta t^m$  start with large values then they gradually become smaller and smaller as k increases. On the other hand, for the first two cases when k=50, 100,  $\Delta t^l$  and  $\Delta t^m$  are small when l and m are small, then they increase to reach a peak, then they are gradually decreasing. However, in the third case when k=200, both  $\Delta t^l$  and  $\Delta t^m$  start with large values then they gradually become smaller and smaller as l and  $\Delta t^m$  start with large values then they gradually become smaller and smaller and  $\Delta t^n$  are small when l and m are small, then they increase to reach a peak, then they are gradually decreasing. However, in the third case when k=200, both  $\Delta t^l$  and  $\Delta t^m$  start with large values then they gradually become smaller and smaller and m are s and m, respectively, increases.



**Fig. 1.** Adaptive time-step sizes,  $\Delta t^l$  and  $\Delta t^m$  against the number of steps of the outer loop k and the number of the inner loops l and m: Case 1 (k=50).

Variations of saturation of the heterogenous permeability are shown in Fig. 4. It is noteworthy that the distribution for water saturation is discontinuous due to the high heterogeneity. Thus, for example, we may note higher water saturation at higher permeability regions. Similarly, one may observe the discon-



**Fig. 2.** Adaptive time-step sizes,  $\Delta t^l$  and  $\Delta t^m$  against the number of steps of the outer loop k and the number of the inner loops l and m: Case 2 (k=100).



**Fig. 3.** Adaptive time-step sizes,  $\Delta t^l$  and  $\Delta t^m$  against the number of steps of the outer loop k and the number of the inner loops l and m: Case 3 (k=200).

tinuity in the nanoparticles concentration as illustrated in Fig. 5. The contrast of the nanoparticles concentration distribution arisen from the heterogeneity of the medium can also be noted here. Moreover, the contours of the deposited nanoparticles concentration are shown in Fig. 6. The behavior of the entrapped nanoparticles concentration  $c_{s2}$  is very similar to the behavior of  $c_{s1}$  because of the similarity in their governing equation.



Fig. 4. Variations of saturation of the real heterogenous permeability case.

# 6 Conclusions

In this paper, we introduce an efficient time-stepping scheme with adaptive timestep sizes based on the CFL calculation. Hence, we have calculated the  $CFL_{s,x}$ ,  $CFL_{s,y}$ ,  $CFL_{c,x}$  and  $CFL_{c,y}$  at each substep and check if the CFL condition is satisfied (i.e. CFL < 1). We applied this scheme with the IMES-IMC scheme to simulate the problem of nanoparticles transport in two-phase flow in porous media. The model consists of five differential equations, namely, pressure, saturation, nanoparticles concentration, deposited nanoparticles concentration on the pore-walls, and entrapped nanoparticles concentration in pore–throats. The capillary pressure is linearized and used to couple the pressure and the saturation equations. Then, the saturation equation is solved explicitly to update the saturation at each step. Therefore, the nanoparticles concentration equation is treated implicitly. Finally, the deposited nanoparticles concentration on the



Fig. 5. Variations of nanoparticles concentration of the real heterogenous permeability case.

pore-walls equation, and entrapped nanoparticles concentration in pore-throats equation are solved implicitly. The CCFD method was used to discretize the governing equations spatially. In order to show the efficiency of the proposed scheme, we presented some numerical experiments. The outer pressure time-step size was selected then the inner ones, namely, the saturation subtime-step and the concentration subtime-step were calculated and adaptive by the CFL condition. We presented three cases with different values of the outer pressure time-step. Moreover, distributions of water saturation, nanoparticles concentration, and deposited nanoparticles concentration on pore-wall are shown in graphs.

# References

- 1. Chen, Z., Huan, G., Ma, Y.: Computational methods for multiphase flows in porous media. SIAM Comp. Sc. Eng., Philadelphia (2006).
- Young, L.C., Stephenson, R.E.: A generalized compositional approach for reservoir simulation: SPE Journal 23, pp. 727-742 (1983).
- 3. Kou, J., Sun, S.: A new treatment of capillarity to improve the stability of IMPES two-phase flow formulation. Computers & Fluids 39, 1923-1931 (2010).
- Belytschko, T., Lu, Y.Y.: Convergence and stability analyses of multi-time step algorithm for parabolic systems. Comp. Meth. App. Mech. Eng. 102, 179-198 (1993).
- Gravouil A., Combescure, A.: Multi-time-step explicit—implicit method for nonlinear structural dynamics. Int. J. Num. Meth. Eng. 50, 199-225 (2000).



Fig. 6. Variations of deposited nanoparticles concentration of the real heterogenous permeability case.

- Klisinski, M.: Inconsistency errors of constant velocity multi-time step integration algorithms. Comp. Ass. Mech. Eng. Sc. 8, 121-139 (2001).
- Bhallamudi, S.M., Panday, S., Huyakorn, P.S.: Sub-timing in fluid flow and transport simulations, Adv. Water Res. 26, 477-489 (2003).
- Park, Y.J., Sudicky, E.A., Panday, S., Sykes, J.F., Guvanasen V.: Application of implicit sub-time stepping to simulate flow and transport in fractured porous media. Adv. Water Res. 31, 995-1003 (2008).
- Singh, V. and S. M. Bhallamudi, S.M.: Complete hydrodynamic border-strip irrigation model. J. Irrig. Drain. Eng. 122, 189-197 (1996).
- Smolinski, P. Belytschko, T. Neal, M.: Multi-time-step integration using nodal partitioning," Int. J. Num. Meth. Eng. 26, 349-359 (1988).
- 11. Hoteit, H., Firoozabadi, A.: Numerical modeling of two-phase flow in heterogeneous permeable media with different capillarity pressures. Adv. Water Res. 31, 56-73 (2008).
- Gruesbeck, C., Collins, R.E.: Entrainment and deposition of fines particles in porous media. SPE J. 24, 847-856 (1982).
- 13. Ju, B. and Fan, T.: Experimental study and mathematical model of nanoparticle transport in porous media. Powder Tech.192, 195-202 (2009).
- 14. Liu, X.H., Civian, F.: Formation damage and skin factor due to filter cake formation and fines migration in the Near-Wellbore Region. SPE-27364, In: SPE Symposium on Formation Damage Control, Lafayette, Louisiana (1994).
- El-Amin, M.F., Salama, A., Sun, S.: Modeling and simulation of nanoparticles transport in a two-phase flow in porous media. SPE-154972, In: SPE International Oilfield Nanotechnology Conference and Exhibition, Noordwijk, The Netherlands (2012).
- 16. El-Amin, M.F., Sun, S., Salama, A.: Modeling and simulation of nanoparticle transport in multiphase flows in porous media: CO<sub>2</sub> sequestration. SPE-163089, In: Mathematical Methods in Fluid Dynamics and Simulation of Giant Oil and Gas Reservoirs (2012).
- El-Amin, M.F., Sun, S., Salama, A.: Enhanced oil recovery by nanoparticles injection: modeling and simulation. SPE-164333, In: SPE Middle East Oil and Gas Show and Exhibition held in Manama, Bahrain (2013).
- El-Amin, M.F., Salama, A., Sun, S.: Numerical and dimensional analysis of nanoparticles transport with two-phase flow in porous media. J. Pet. Sc. Eng. 128, 53-64 (2015).
- Salama, A., Negara, A., El-Amin, M.F., Sun, S.: Numerical investigation of nanoparticles transport in anisotropic porous media. J. Cont. Hydr. 181, 114–130 (2015).
- Chen, H., Di, Q., Ye, F., Gu, C., Zhang, J.: Numerical simulation of drag reduction effects by hydrophobic nanoparticles adsorption method in water flooding processes. J. Natural Gas Sc. Eng. 35, 1261-1269 (2016).
- Chen, M.H., Salama, A., El-Amin, M.F.: Numerical aspects related to the dynamic update of anisotropic permeability field during the transport of nanoparticles in the subsurface. Procedia Comp. Sc. 80, 1382-1391 (2016).
- El-Amin, M.F., Kou, J., Salama, A. and Sun, S.: An iterative implicit scheme for nanoparticles transport with two-Phase flow in porous media. Procedia Comp. Sc. 80, 1344-1353 (2016).
- Al-Dhafeeri, A.M., Nasr-El-Din, H.A.: Characteristics of high-permeability zones using core analysis, and production logging data. J. Pet. Sc. Eng. 55, pp. 18–36 (2007).