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Numerical Simulation Of Pore Fluid Flow And Fine Sediment Infiltration Into The Riverbed

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The riverbed embodies an important ecotone for many organisms. It is also an interface between groundwater and surface water, two systems that feature numerous distinctions. The riverbed therefore exhibits many physical and biochemical gradients e.g. flow velocity, temperature, oxygen or nutrient concentration. If the riverbed however becomes clogged through input and deposition of fine sediments, its porosity and permeability decrease leads to reduced interconnectivity between both neighboring systems and eventually reduction of the suitability of the riverbed as a habitat for organisms. Reasons for fine sediment infiltration are high inputs of fine sediments from surface runoff or rainwater retention basins and continuous unnatural low flow velocities typically found in regulated rivers. The objective of our current research is to develop a model for the determination of the factors controlling fine sediment infiltration into the riverbed and their quantitative impact on fine sediment infiltration rates, reduction of riverbed porosity, and permeability. To do so, we use several numerical modeling techniques including the popular lattice Boltzmann method.

**INTRODUCTION**

The hyporheic zone (or river bed) plays an important role in the field of integrated river management, due to its relevance as an ecotope for many organisms and interface between surface water and ground water. The suitability of the river bed as a habitat for organisms and exchange medium for oxygen and nutrients between ground and surface water strongly depends on existing cavity systems within riverbed sediments. These systems can become blocked if fine sediments infiltrate and deposit within the cavities, thereby reducing the porosity and permeability of the river bed. Fine sediments are defined to be sediment particles with a characteristic diameter $d_c < 2$ mm. Reasons of fine sediment infiltration are high inputs of fine sediments from surface runoff or rainwater retention basins and continuous unnatural low flow velocities typically found in regulated rivers. The objective of our current research is to develop a model for the systematic determination of the factors controlling fine sediment infiltration into the riverbed and their quantitative impact on fine sediment infiltration rates, reduction of riverbed porosity, and permeability. To do so, we run through a set of modeling steps. First, we...
use digitized coarse sediment particles ($d_c > 2$ mm) to generate a riverbed pore structure (or packing). In a second step we use this pore structure as internal solid boundaries for pore-scale lattice Boltzmann (LB) simulations to calculate pore water flow and fine sediment transport. Thereby, fine sediment particle motion is resolved using a Lagrangian particle tracker technique. Fine particles are able to collide with the packing structure which can lead to deposition of fine particles within the pore structure under certain conditions. Deposited fine particles are converted to internal solid boundaries, which enables dynamic pore flow patterns as an answer to advancing pore structure clogging. The pore structure can be exported during the simulation and structure properties like porosity and permeability can be evaluated in a post-processing step. For LB simulations we use the parallel high-performance-computing framework waLBerla developed by the Chair for System Simulation, University of Erlangen-Nürnberg, Germany [1]. Due to a high resolution ($dx \sim 0.1 – 1$ mm) and large dimensions ($10^9 – 10^{11}$ lattice cells) of the 3D-LB model domain, simulations run at the Jülich Supercomputing Centre (JSC), Jülich, Germany. In order to use the existing computer architecture efficiently, simulation algorithms need to fulfill high parallelization standards.

LATTICE-BOLTZMANN MODEL

The lattice Boltzmann method (LBM) became a popular method for hydrodynamic modeling especially due to its simplicity, accuracy, (parallel) efficiency, and straightforward resolution of complex boundaries and multi-phase flows. Historically, it originated from the lattice gas cellular automata (LGCA) [2], but it was shown that the lattice Boltzmann equation can be directly derived by discretizing the Boltzmann equation [3][4]. It has also been shown that the LBM is equivalent to an explicit, first-order in time, second-order in space finite difference approximation of the incompressible Navier-Stokes (NS) equations [5]. With its kinetic origin, the Boltzmann equation, some features of the LBM are significantly different from conventional computational fluid dynamics (CFD) methods based on direct discretization of the NS equations. The macroscopic quantities, such as fluid density and fluid velocity, are obtained by solving the kinetic equation for the particle distribution function $f$ and evaluating the hydrodynamic moments of the distribution function.

The most popular LBM model in flow simulations is the lattice Bhatnagar-Gross-Krook (LBGK) model due to its simplicity. However, the LBGK encounters numerical instability [6] and inaccuracy in implementing boundary conditions. The viscosity dependent boundary conditions pose a severe problem especially for simulating flow through porous media [7] because the permeability becomes viscosity dependent, while it should be a characteristic of the physical properties of porous medium alone. The multi-relaxation-time (MRT) LBM, which we use in our simulations, has proved to reduce these deficiencies [8][9], by separating the relaxation times and allowing to improve numerical stability by tuning different relaxation times individually.
Multi-relaxation-time LBM

The MRT, like any other LBM, is based on three components that discretize the continuous Boltzmann equation. The first component is a discrete phase space. It is defined by a regular lattice in $D$ dimensions and a constant distance between the lattice nodes, the lattice spacing $\delta_x$.

Each lattice node $x_k \in \delta_x \mathbb{Z}^d$ is connected with some of its neighbors by a finite set of symmetric discrete velocities $\{e_{|\alpha|} \mid \alpha = 0, 1, \ldots, N\}$. The second component is a collision matrix $S$ and $B$ equilibrium distribution functions (with $B = N + 1$ assuming that the velocity set has a zero velocity component i.e. $e_0 \equiv 0$, otherwise $B = N$):

$$\{f^{(eq)}_{\alpha} \mid \alpha = 0, 1, \ldots, N\}.$$  

The equilibrium distribution functions are functions of the local conserved quantities. The third component is an evolution equation of the distribution functions in discrete time $t_n \in \delta_t \mathbb{N}_0$:

$$f(x_k + e_\delta, t_n + \delta_t) - f(x_k, t_n) = S[f^{(eq)}(x_k, t_n) - f(x_k, t_n)],$$  

or equivalently

$$f(x_k + e_\delta, t_n + \delta_t) - f(x_k, t_n) = M^{-1} \hat{S} \{m^{(eq)}(x_k, t_n) - m(x_k, t_n)\},$$  

where $f, m,$ and $m^{(eq)}$ are $B$-dimensional (column) vectors. The relaxation matrix $\hat{S} = M \cdot S \cdot M^{-1}$ is a $B \times B$ diagonal matrix. The transformation matrix $M$ relates the distribution functions represented by $f \in \mathbb{V} = \mathbb{R}^B$ to their velocity moments represented by $m \in \mathbb{M} = \mathbb{R}^B$. $M$ is constructed from the monomials of the discrete velocity components via the Gram-Schmidt orthogonalization procedure [10][6]. The relaxation matrix $\hat{S}$ is diagonal in moment space $\mathbb{M}$:

$$\hat{S} = \text{diag}(0, s_1, s_2, 0, s_4, 0, s_5, s_7, s_9, s_{10}, s_{11}, s_{13}, s_{14}, s_{15}, s_{16}, s_{17}, s_{18}, s_{19}),$$  

$$= \text{diag}(0, s_e, s_e, 0, s_q, 0, s_q, s_q, s_q, s_q, s_q, s_q, s_q, s_q, s_q, s_q, s_m, s_m, s_m).$$  

(3)

The viscosity of the D3Q19 model is

$$\nu = \frac{1}{3} \left( \frac{1}{s_e} - \frac{1}{2} \right)$$  

and the speed of sound $c_s = 1/\sqrt{3}$. If all of the relaxation rates, $\{s_\alpha \mid \alpha = 0, 1, \ldots, B\}$, are set to $1/\tau$, thus, $S = \tau^{-1} \cdot I$, with $I$ being the $19 \times 19$ identity matrix, then the model simplifies to a single-relaxation-time (SRT) LBGK model. If the relaxation rates are chosen according to Eq. (5), the model is equivalent to the two-relaxation-times (TRT) model described in [8].

$$s_e = s_e = s_n = s_v = \frac{1}{\tau} \quad s_m = s_q = \frac{8}{3} \left( \frac{2-s_n}{8-s_n} \right).$$  

(5)

The TRT is based on Chapman-Enskog analysis and treats even-order and odd-order modes with two different relaxation rates, $s_v$ and $s_q$, respectively.

Although the LBGK is still the most popular LBM for flow simulations, both, MRT and TRT show improvements in terms of accuracy and numerical stability, especially in porous medium.
flow simulations. Anyway, due to a more complex relaxation procedure, the MRT could be about 15% slower than the LBGK in terms of lattice site updates per second (LUPS) \[11\], while TRT performance is equal to the LBGK \[1\]. Since our simulations run on massively parallel computers like JUQUEEN, computational performance is a major criterion. Therefore, we use the TRT collision scheme in our current LBM simulations.

**FINE SEDIMENT PARTICLES**

The process of fine sediment deposition exposed to current can be divided into three stages: particle motion, collision, and deposition. Each stage is handled in a self-contained computational unit.

**Particle Motion**

Three-dimensional turbulent transport of suspended sediment in a fluid can be described by the following equation

\[
\frac{\partial c}{\partial t} = \frac{\partial}{\partial x_i} \left( D_t \frac{\partial c}{\partial x_i} \right) - u_i \frac{\partial c}{\partial x_i},
\]

which emerges from the advection-diffusion equation by applying simplifications due to the assumption of incompressible flow and neglecting molecular diffusion. In Eq. (6), \( c \) is the solute concentration, \( D_t \) the turbulent diffusion coefficient, and \( u \) is the flow velocity. If the transport is governed by Eq. (6), the Lagrangian movement of a particle can be described by:

\[
r(t + \Delta t) = r(t) + (u + \nabla \cdot D_t) \Delta t ,
\]

where \( r \) is the three-dimensional particle position vector \((x, y, z)^T\), and \( \nabla \cdot \) is divergence. Eq. (7) is an explicit Euler scheme for the integration of the equations for the particle movement with an approximation of \( O(\Delta t) \). Other numerical schemes, like predictor-corrector or Runge-Kutta, exist and show better approximations in the order of \( O(\Delta t^2) \) and higher. Anyway, for small time steps \( \Delta t \), Eq. (7) still leads to acceptable results. For situations where turbulent diffusion can be neglected, e.g. laminar flow conditions, Eq. (7) can be simplified to

\[
x(t + \Delta t) = x(t) + u_x \Delta t ,
\]

\[
y(t + \Delta t) = y(t) + u_y \Delta t ,
\]

\[
z(t + \Delta t) = z(t) + (u_x - u_z) \Delta t ,
\]

with the particle sedimentation velocity \( u_z \) (typically along the vertical \( z \)-axis). It should be mentioned here that fine particles are not treated as inertial particles. Hence, they are not exposed to acceleration. They are rather modeled as point masses without any spatial extent. This simplification however, is valid for particles with a small size in the order of the lattice.
spacing $\delta_x$, which is satisfied for fine sediment particles with a characteristic diameter $d_c$ of $d_c < \delta_x$. To retain mass conservation during the deposition step, fine particle diameters should be chosen to $\sqrt[6]{6/\pi} \delta_x$, if they are assumed as perfect spheres.

**Collision Detection**

Figure 1 shows the three different possible cases that the collision detection has to deal with. Because fine particles can travel several lattice cells into or even across coarse particles during a single time step, it is necessary to search for fine-coarse particle collisions along the trace of each fine particle at every time step, which can be very time consuming for many particles.

The collision detection is separated into two subsequent steps. First, the fine particle trace, which is described by real world coordinates, i.e. floating-point numbers, has to be mapped onto the LBM lattice, which is described by integer values. This process is called voxelization as lattice cells are also named voxels. Most voxelization algorithms only treat integer-based geometries [12] as input since they were developed for scan-conversion purposes. Hence, they are not applicable for our purposes, as we are dealing with floating-point input geometries. Our collision detection algorithm transforms lattice cells to geometric objects based on floating point numbers and identifies trace-object intersections based on a standard ray-tracing method. This method leads to a unique voxel representation of each fine particle trace. The second step of the collision detection algorithm detects coarse particle boundary cells along the voxelized trace starting at the last fine particle position $r(t = i - 1)$. The firstly detected boundary cell is the collision cell. The preceding cell, meaning the fluid cell neighboring the collision cell, is provided during the deposition step as the preferred fine particle deposition site.

![Figure 1. Three possible fine-coarse particle collision types. The fine particle velocity $u(t = i)$ increases from left (a) to right (c).](image)
Deposition Step

The deposition step converts a moving fine particle to an internal solid boundary condition in the LBM simulation, thereby modeling the riverbed clogging process. Hence, dynamic interstitial flow path patterns can be simulated, which is a superior objective of our research. The preferred deposition site is determined during the collision step, but is not necessarily the actual deposition site. Neighboring sites can also be chosen in order to achieve a more realistic deposition pattern. In the current implementation all fine-coarse particle collisions lead to deposition of fine particles at the preferred deposition site. Although this assumption may only be valid for very special hydraulic situations, resulting deposition patterns can be used as indicators for further investigations with more complex deposition models. Moreover, the deposition step does not consume any computational resources, since no additional calculations are necessary.

SIMULATIONS WITH FINE SEDIMENT INFILTRATION

Model Schematization

A packing structure, representing a virtual riverbed, has been generated using a digital packing algorithm [13]. The domain size is $100^3$ lattice cells. The lattice spacing $\delta_x$ is 1 mm in physical units, and the time step $\delta_t = 0.02$ s. The flow velocity at the inlet boundary is $u_x = 0.1$ m/s for lattice cells above packing height and $u_x = 0.0$ m/s for cells below packing height. The outlet boundary is placed on the y-z-plane opposite to the inlet boundary, thereby introducing current along the x-axis. The top boundary also imposes a flow velocity $u_y = 0.1$ m/s on the domain. The bottom boundary is modeled via a no-slip condition; all other boundaries are periodic boundaries. Fine particles with $d_c = \sqrt[3]{6/\pi} \delta_x = 0.62$ mm enter the domain normally distributed from the top and inlet boundary. Due to a predefined sedimentation velocity of $u_d = 0.1$ m/s, particles sink to the bottom and collide/deposit on the coarse particle packing structure. The simulation runs for 3000 time steps, which is one minute in real time. In simulation (I) the particle addition rate is 5 and in simulation (II) 500 particles per second, respectively.

Results

The initial packing structure is illustrated in Figure 2a-b and the results of two simulations are illustrated in Figure 2c-f. The porosity $n$, which is defined as the ratio of pore volume to total sample volume decreases during the simulation from 0.5 at the initial state to 0.498 and 0.442, respectively. Although 30,000 particles were simulated in (II), for which the collision detection algorithm has to be evaluated at every time step individually, no remarkable increase in computation time could be noticed.
DISCUSSION

Our results show that fine particle movement in a complex flow field can be simulated even with high numbers of fine particles. The LBM flow model confirms promising features of the TRT collision model, such as good accuracy of the results, even in complex porous geometries, and very good parallel performance. Fine-coarse particle collisions are reliably detected even with high fine particle sedimentation velocities. External riverbed clogging (deposition of fine sediments on the topmost layer of the riverbed) can be simulated for a limited number of fine particles. Anyway, the results of simulation (II) show unnatural tower-like fine particle deposition patterns. These patterns result from the deposition step, which assumes that a fine particle deposits exactly where it collides with the packing structure, i.e. at the preferred deposition site provided by the collision detection step. The penetration of the upper part of the packing structure will become unlikely if horizontal components of the flow velocity increase. Fine particles will more often deposit after horizontal collisions which may lead to external clogging in an early infiltration state.

Figure 2. Packing structures (a, c, e) and cross sections (b, d, f) made of even spheres. Grey particles represent coarse sediment particles and black particles represent fine sediment particles. Volume above packing height has been removed due to simplicity. (a-b): no fine particle deposits, $n = 0.500$. (c-d): after 1 minute of fine particle rain (addition rate: 5 particles per second), $n = 0.498$. (e-f): after 1 minute of particle rain (addition rate: 500 particles per second), $n = 0.442$. 

CONCLUSION

The collision detection showed good performance in terms of computational efficiency. To avoid tower-like deposit growth, the actual deposition location of fine particles should be chosen more carefully. The effect of horizontal deposition can be avoided by applying different deposition treatments for horizontal and vertical collision. The implementation of a re-mobilization step, preferably based on an LBM-flavored force balance, is still pending.

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