

# HPC Mobile Platform for Solving Oil Recovery Problem

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**Abstract:** The paper describes applying of mobile computational tools to numerical solving of full value industrial problems. As an example, we used surfactant/polymer flooding problem with thermal effects. The problem solved by using different graphics processing units including the GPUs of mobile devices. Parallel implementation of numerical algorithm was launched on the following devices: NVidia GeForce 770, NVidia Tesla K20 and mobile GPU NVidia Tegra K1. We developed mobile application implementing the algorithm. The application was tested and compared with desktop GPUs of same microarchitecture. Results of the tests shows that calculation on the mobile devices gives the same computation efficiency as desktop GPU with average characteristics.

## 1 INTRODUCTION

Using mobile devices in computation of physical simulations are not yet common. However, the large number of hardware demanding applications on mobile devices proves that these devices have sufficient computing power.

Implementation of some grid systems' nodes as mobile devices, for example, considered in (Ketan, 2013; Phan, 2002). A clear example of such system developed up to industrial scales is BOINC (boinc.berkeley.edu). Nevertheless, such systems do not consider heterogeneity of mobile processors. In such applications only CPU kernel of the processor is used.

However, GPUs of mobile devices are quite suitable for resource intensive computations. They are widely used for image recognition problems (Singhal, 2010; Wang, 2013). Also, (Cheng, 2011) gives an idea of general-purpose computation on mobile graphics processors.

High performance parallel computing using CUDA has been attracting many researchers in various disciplines, including computational fluid dynamics (Tolke, 2008; Micikevicius; Thibault, 2009; Kelly, 2013). This work presents applying and testing mobile GPU for simulation of surfactant/polymer flooding (SPF) process, which is taken as an example of complex industrial problem. Polymer/surfactant flooding is one of the effective chemical enhanced oil recovery (EOR) methods. Polymer increases

viscosity of a water thereby improving the mobility ratio and increasing the recovery efficiency. Primary benefit of polymer flooding is to improve sweep efficiency and acceleration of oil production (Lake, 1983; Sorbie, 1991). Surfactant flooding method involves addition of surface-active agents or surfactants to the injected water. Surfactants reduces the interfacial tension between oil and water within reservoir, reduce the residual oil saturation and improve displacement efficiency (Babalyan, 1983).

For test analysis results of mobile GPU computation were compared with two PC NVidia GPUs on Kepler microarchitecture. The reason of the technology selection is that available at the moment computational mobile GPU NVidia Tegra K1 is also has this microarchitecture. First GPU used for comparison is one of the most forward computational processors NVidia Tesla K20 and the second one is common GPU NVidia GeForce 770 with average features and performance.

Mobile application implementing described simulation is presented at the end of the paper after tests and analysis. Further, it may be developed for computing any continuous media simulations and also using the mobile GPU as nodes of large heterogeneous distributed systems.

This paper is structured as follows. In section 2, we consider the mathematical model of SPF problem and implementation of numerical solution. In section 3, we discuss parallel algorithm for solving SPF problem using CUDA and calculation time tests on different devices.

## 2 MATHEMATICAL MODEL OF SPF AND ITS NUMERICAL SOLUTION

The mathematical model of two-phase flow in porous media has following assumptions:

- incompressible flow;
- gravitational forces and capillary effects is neglected;

- two-phase flow (water, oil) obeys Darcy's law.

Mass conservation equations and velocities for each phase can be written as follows:

$$m \frac{\partial S_w}{\partial t} + \text{div}(\mathbf{v}_w) = q_1 \quad (1)$$

$$m \frac{\partial S_o}{\partial t} + \text{div}(\mathbf{v}_o) = q_2 \quad (2)$$

$$S_w + S_o = 1$$

$$\mathbf{v}_i = -K_0 \frac{f_i(s)}{\mu_i} \nabla P, \quad i = w, o \quad (3)$$

where  $m$  – porosity,  $S_w, S_o$  – water and oil saturations,  $q_1, q_2$  – source or sink,  $\mathbf{v}_w, \mathbf{v}_o$  – velocity of the water and oil phases,  $f_i(s), \mu_i$  – relative permeability and viscosity for phase  $i$ ,  $K_0$  – permeability tensor.

Polymer, surfactant, salt and heat transport equations are given by (Babalyan, 1983):

$$m \frac{\partial}{\partial t} (c_p S_w) + \frac{\partial a_p}{\partial t} + \text{div}(\mathbf{v}_w c_p) = \text{div}(m D_{pw} S_w \nabla c_p) \quad (4)$$

$$m \frac{\partial}{\partial t} (c_{sw} S_w + c_{so} S_o) + \frac{\partial a_{surf}}{\partial t} + \text{div}(\mathbf{v}_w c_{sw} + \mathbf{v}_o c_{so}) = \text{div}(m D_{sw} S_w \nabla c_{sw} + m D_{so} S_o \nabla c_{so}) \quad (5)$$

$$m \frac{\partial}{\partial t} (c_s S_w) + \text{div}(\mathbf{v}_w c_s) = 0 \quad (6)$$

$$\frac{\partial}{\partial t} [(1 - m) C_r \rho_r + m(C_w S_w \rho_w + C_o S_o \rho_o) T] + \text{div}(\rho_w C_w \mathbf{v}_w) + \text{div}(\rho_o C_o \mathbf{v}_o) = \text{div}[(1 - m) \lambda_0 + m(\lambda_1 S_w + \lambda_2 S_o)] \nabla T \quad (7)$$

where  $c_p, c_s$  – polymer and salt concentrations in aqueous phase,  $c_{sw}, c_{so}$  – surfactant concentration in water and oleic phases,  $a_p, a_{surf}$  – polymer and surfactant adsorption functions,  $D_{pw}, D_{sw}, D_{so}$  – polymer and surfactant diffusion coefficients,  $C_w, C_o, C_r$  – specific heat of water, oil and rock,  $\rho_w, \rho_o, \rho_r$  – density of water, oil and rock,  $\lambda_0, \lambda_1, \lambda_2$  – coefficients of thermal conductivity.

Initial conditions:

$$\begin{aligned} S_w|_{t=0} &= S_{w0}, \\ c_{pw}|_{t=0} &= c_{p0}, \quad c_s|_{t=0} = c_{s0}, \\ c_{sw}|_{t=0} &= c_{sw0}, \quad c_{so}|_{t=0} = c_{so0}, \\ T|_{t=0} &= T_p \\ a_{surf0}|_{t=0} &= a_{surf0}, \quad a_p|_{t=0} = a_{p0}, \end{aligned} \quad (8)$$

Boundary conditions:

$$\begin{aligned} \frac{\partial S_w}{\partial n} \Big|_{\partial \Omega} &= 0; \quad \frac{\partial P}{\partial n} \Big|_{\partial \Omega} = 0; \quad \frac{\partial T}{\partial n} \Big|_{\partial \Omega} = 0; \\ \frac{\partial c_{pw}}{\partial n} \Big|_{\partial \Omega} &= 0; \quad \frac{\partial c_{sw}}{\partial n} \Big|_{\partial \Omega} = 0; \quad \frac{\partial c_s}{\partial n} \Big|_{\partial \Omega} = 0; \end{aligned} \quad (9)$$

We used the following viscosity dependence on injected reagent concentrations and temperature: (Flory-Huggins, 1953):

$$\mu_a = \mu_w [1 + (\gamma_1 c_p + \gamma_2 c_p^2 + \gamma_3 c_{sw} + \gamma_4 c_{sw}^2) c_s^{\gamma_5} - \gamma_6 (T - T_p)] \quad (10)$$

$$\mu_o = \mu_{o0} [1 - \gamma_7 (T - T_p)] \quad (11)$$

where  $\gamma_1, \gamma_2, \gamma_3, \gamma_4, \gamma_5, \gamma_6, \gamma_7$  – constants.  $\mu_{o0}$  – initial viscosity of oleic phase,  $T_p$  – reservoir temperature. The imbibition relative permeability curve for water/oil flow is given by

$$f_w(S_w) = S_w^{3.5}; \quad f_o(S_w) = (1 - S_w)^{3.5}$$

The adsorbed concentration of polymer is a function of polymer concentration, given by:

$$a = \frac{b c_p}{1 + b c_p}$$

where  $b$  – Langmuir constant.

The numerical formulation of equations (1)-(9) based on finite difference method and explicit scheme. The explicit schemes are naturally parallelizable using the GPUs. Numerical realization and results of numerical experiments was proposed by the authors in (Danaev, 2015; Ahmed-Zaki, 2015).

## 3 TESTING RESULTS OF PARALLEL ALGORITHM USING CUDA TECHNOLOGY

For testing of the SPF problem we used devices with NVIDIA Tesla K20, GeForce GTX 770 and Tegra K1 video cards with Kepler microarchitecture (www.nvidia.com). For further testing, we compared the characteristics of the devices. All characteristics affect the performance of programs calculation (Table 1).

Table 1: Description of the NVIDIA Kepler architecture video cards.

	Tesla K20 GK110	GeForce GTX 770 GK104	Tegra K1 GK20A
Cores	2496	1536	192
Clock Speed [MHz]	732	1046	950
Memory bandwidth [GB/s]	208	224	17
Memory (max) [GB]	5	2	1
TDP(Watt)	225	230	10
Compute Capability	3.5	3.0	3.0
Theoretical performance single (GFLOPS)	3524	3213	365
Theoretical performance double (GFLOPS)	1175	134	290

Let NX, NY and NZ to be respectively a number of nodes in the x, y, and z directions of the computational domain. Three-dimensional area with size NX x NY x NZ will be presented as one-dimensional array of size NX x NY x NZ on CPU side. One-dimensional representation of data in global memory is also used on the GPU. Data distribution in the device memory is performed once at the beginning of calculation. On GPU data stored in the shared memory of processor kernel. Computational threads of each block copies data from global memory to the shared memory. The calculation is performed in the threads using that data. Thereafter, the calculation results are written back into the global memory before finishing the kernel function. To obtain the benefit of using shared memory arithmetic operations in the core must be complex enough to compensate the cost of copying the data. One way to achieve this is to increase size of the block. Working with the shared memory we must consider that it has restriction of 16 Kbytes.

Table 2: Memory required for array initialization.

Grid size	Required memory (Gb)
32x32x32	~0.005
64x64x64	~0.042
128x128x128	~0.344
256x256x256	~2.750

In the numerical algorithm of the solution, each computational node needed 176B memory space. So implementing memory allocation to whole grid required particular amount of free space (Table 2). Tegra K1 could not launch program computing the

problem on 256x256x256 grid. It can be explained by the fact that the RAM of used mobile device is only 1GB and initialization arrays did not fit to it.

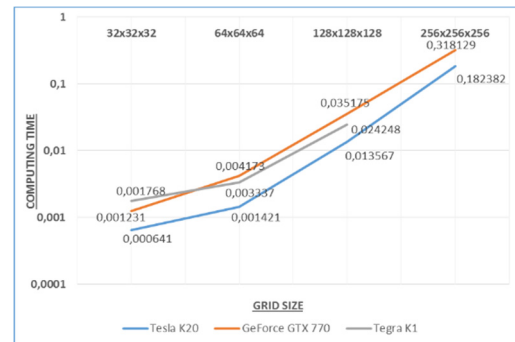


Figure 1: Program calculation time on different devices.

Tests show that Tegra K1 graphic card computes on the same level with other tested ones. For example, on a 64x64x64 grid the mobile device performs the calculation faster than the GeForce GTX 770. In this case, we cannot find an excuse of architecture difference, since the microarchitecture of processors are the same. It can also be explained by the fact that algorithm minimizes addressing to global memory. As we can see in Table 1 memory bandwidth is a weak point of Tegra K1 and we are reducing its usage as much as possible. Tesla K20 has three times lesser calculation time since its characteristics significantly exceeds the remaining cards (figure 1).

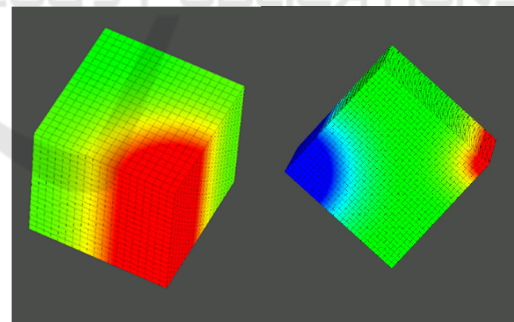


Figure 2: Distribution of water saturation and pressure.

The results of parallel computational experiments conducted on the tablet presented on figures 2-3. Including OpenGL visualization it forms mobile application that can be used to calculate main technological parameters of oil recovery.

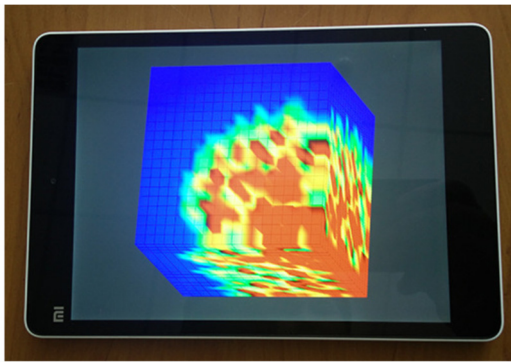


Figure 3: Demonstration of the mobile application results.

## 4 CONCLUSIONS

The paper describes usability of mobile devices to GPGPU calculation of industrial scale problem. As the problem we took a mathematical model of oil displacement process by polymer/surfactant injection. We presented the problem as an example of complex industrial simulation. The problem was solved by explicit numerical method because it well suits to GPGPU parallelization.

The main steps of the numerical algorithm are implemented with separate CUDA kernel functions by using shared memory. The reason is that mobile device graphics card has an architecture that is adverse to only global memory algorithms. This is due to the fact that device has combined CPU and GPU RAM.

By testing calculation time of the program on different grids, we will notice that the mobile device with the Tegra K1 video card, not much inferior to device with the Tesla K20 video card and practically equal to GeForce GTX 770. This suggests that the complex hydrodynamic problems can run wherever there is a mobile device with a video card that supports CUDA technology.

Engineers can use presented mobile application for planning and analyze of oil recovery on real oil fields. Our future work will focus on the use of graphics cards power to calculate programs at a time on several mobile devices. We also plan to expand our code for heterogeneous computing.

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