# 2.<sup>st</sup> Task in *HPC-I*

Deadline: Jan 5, 2024, 11:59pm

### PDE solvers using CUDA

We are going to investigate and implement various solvers for linear systems of equations on GPU.

One long-distance goal consists in a potential incorporation as GPU-solver for (non-linear) systems of equations in the CARP context

Download first:

- the simple conjugate gradients solver<sup>1</sup> together with input data<sup>2</sup> (2.9 GB),
- the NVIDIA example<sup>3</sup> for direct solvers,
- my book<sup>4</sup> as with some algorithms in  $\S6$ .

**0. Given Code** The directory of the cg solver contains two implementations:

- (A)  $cg\_example.cpp$ : The cg solver (git<sup>5</sup>) with ICC-preconditioning by NVIDIA from its library samples<sup>6</sup>.
  - See lines 2,3 how to compile and start the code, or run make special
  - Check the cg implementation by NVIDIA, I suspect an algorithmic error in lines 367-373, see also Alg. 6.19 in the book.
- (B) *main.cpp*: My cg for further work.
  - make run
  - ./main.NVCC\_ data/square\_100\_n uses other input files with  $n \in [0, 7]$ .
  - cg with diagonal preconditioning is implemented in the CPU as well as in the GPU part.

w = r,

 $w = D^{-1}r,$ 

 $LDL^Tw = r$ 

We should have 3 variants for preconditioning in (B) at the end of 2.:

- identity (no preconditioning)
- diagonal pc
- IC (incomplete<sup>7</sup> Cholesky decomposition<sup>8</sup>)

<sup>&</sup>lt;sup>1</sup>http://imsc.uni-graz.at/haasegu/Download/cg.zip

<sup>&</sup>lt;sup>2</sup>http://imsc.uni-graz.at/haasegu/Download/Math2CPP\_data.zip

<sup>&</sup>lt;sup>3</sup>http://imsc.uni-graz.at/haasegu/Download/cusolver\_examples-main.zip

<sup>&</sup>lt;sup>4</sup>https://imsc.uni-graz.at/haasegu/Lectures/Master\_HPC/textbook.pdf

<sup>&</sup>lt;sup>5</sup>https://github.com/NVIDIA/CUDALibrarySamples/tree/master/cuSPARSE/cg

<sup>&</sup>lt;sup>6</sup>https://github.com/NVIDIA/CUDALibrarySamples

<sup>&</sup>lt;sup>7</sup>https://en.wikipedia.org/wiki/Incomplete\_Cholesky\_factorization

<sup>&</sup>lt;sup>8</sup>https://en.wikipedia.org/wiki/Cholesky\_decomposition

#### 1. Check/compare the cg solvers, add IC to (B)

- (i) Write an additional constructor for CRS\_Matrix\_GPU that reads the date via the *mtx* files from (A).
- (ii) Compare iteration history and run time of the cg algorithms in (A) and (B) on GPU without preconditioning.
- (iii) Check the cg implementation by NVIDIA (A), I suspect an algorithmic error in lines 367-373, see also Alg. 6.19 in the book.
- (iv) Transfer the setup and application of the IC-preconditioning from (A) to (B).
- (v) Compare iteration numbers as well as run time with data from (A) and from (B).

#### 2. Improve/compare the cg solver (B)

 $[\longrightarrow ZR + MK]$ 

Tasks for implementational variants of diagonal preconditioned cg in (B). Use also nsys-ui ./main.NVCC for profiling.

- (vi) Compare run time for unified memory allocation (cudaMallocManaged) versus exclusive device memory allocation (cudaMalloc).
- (vii) Substitute the cuBLAS routines in cg by your own kernel calls and compare the run time. This will require to use file suffix .cu instead of .cpp.
- (viii) Combine computation of  $\underline{r}, \underline{w}$  (,  $\underline{u}$ ) and  $\sigma$  in one kernel call. What about the resulting run time? Take care for register spilling. Is some overlap via different streams possible u vs. the other vectors?
- (ix) Use your own kernel call for the sparse matrix product calculating  $\underline{v}$  (timing regarding the cuSPARSE call) and combine that kernel with the inner product calculation ( $\underline{s}, \underline{v}$ ).

#### **3.** Other iterative solvers and improvements. On basis of your version of (B):

- (x) Can the IC preconditioner be separated into a setup step (memory allocation, pattern of L) and an update step (recalculation of entries in L). This would be interesting for non-linear solvers.
- (xi) Implement a gmres solver, Alg. 6.21 in the book.
  - Notice that dense matrix H is dynamically growing with each iteration.
  - Can the reallocation of H as well as the Givens rotation be overlapped with some other computations GPU?

To be continued (next page)

The html-online material by NVIDIA is more a brief introduction into the user interface than a full description of all function call in a library, see the appropriate pdf-links for the latter one.

- cuSOLVER<sup>9</sup>:
  - Dense and sparse matrices (CSR).
  - QR/LU/Cholesky factorization on (multiple) GPUs.
  - Re-factorization with the same sparsity pattern.
  - Bottleneck: Fill in causes large memory requirements and increased run time.
- cuSPARSE<sup>10</sup>:
  - Supports several sparse formats (incl. CSR) and dense format, see  $\S6.5^{11}$ .
  - BLAS 1-3 functionality for sparse matrices.
    Especially multiplication of two sparse matrices, see cusparseSpGEMM<sup>12</sup> and cusparseSpGEMMreuse<sup>13</sup>.
  - ICC(0) and ILU(0) are implemented for CRS-matrices, see §11 in cuSPARSE-Docu<sup>14</sup> and a preconditioner primer by Nvidia<sup>15</sup>.
  - Re-factorization with the same sparsity pattern might be possible, look for cusparse<t>csrilu02\_analysis() and cusparse<t>csrilu02().
- $AmgX^{16}$ :
  - Algebraic multigrid solver and preconditioner.
  - Krylov solvers (cg, g,res etc.) available.
  - MPI and OpenMP support.
  - Example<sup>17</sup>.
  - Has to be installed separately from the AMG repository<sup>18</sup>.
- $cuDSS^{19}$ 
  - Direct sparse solver library.
  - LU/Cholesky factorization with 3 stages of factorization process: symbolic factorization, numeric factorization, solving step which are controlled via cudssPhase\_t<sup>20</sup> in the cudssExecute()<sup>21</sup> function.
  - API simular to cuSPARSE.
  - Still in development (preview).
  - Has to be installed<sup>22</sup> separately.

<sup>&</sup>lt;sup>9</sup>https://docs.nvidia.com/cuda/cusolver/index.html

<sup>&</sup>lt;sup>10</sup>https://docs.nvidia.com/cuda/cusparse/

<sup>&</sup>lt;sup>11</sup>https://docs.nvidia.com/cuda/cusparse/#sparse-matrix-apis

<sup>&</sup>lt;sup>12</sup>https://docs.nvidia.com/cuda/cusparse/#cusparsespgemm

<sup>&</sup>lt;sup>13</sup>https://docs.nvidia.com/cuda/cusparse/#cusparsespgemmreuse

<sup>&</sup>lt;sup>14</sup>https://docs.nvidia.com/cuda/pdf/CUSPARSE\_Library.pdf

<sup>&</sup>lt;sup>15</sup>https://docs.nvidia.com/cuda/incomplete-lu-cholesky/index.html

<sup>&</sup>lt;sup>16</sup>https://github.com/NVIDIA/AMGX

<sup>&</sup>lt;sup>17</sup>https://developer.nvidia.com/amgx

<sup>&</sup>lt;sup>18</sup>https://github.com/NVIDIA/AMGX

<sup>&</sup>lt;sup>19</sup>https://developer.nvidia.com/cudss

<sup>&</sup>lt;sup>20</sup>https://docs.nvidia.com/cuda/cudss/functions.html#cudsscreate

<sup>&</sup>lt;sup>21</sup>https://docs.nvidia.com/cuda/cudss/functions.html#cudssexecute

<sup>&</sup>lt;sup>22</sup>https://developer.nvidia.com/cudss-downloads

## References

- [HaSh19] Jaegeun Han and Bharatkumar Sharma. Learn CUDA Programming. Packt> (2019); buy<sup>23</sup>; download code<sup>24</sup>, download figures<sup>25</sup>
- [NSLS14] Dan Negrut, Radu Serban, Ang Li and Andrew Seidl. Unified Memory in CUDA 6: A Brief Overview and Related Data Access/Transfer Issues. TR-2014-09<sup>26</sup> (2014)
- [ChAn17] Andrzej Chrzeszczyk and Jacob Anders Matrix computationson the GPUCUBLAS, CUSOLVER and MAGMA by examples. NVIDIA (2017); download book<sup>27</sup>.

Overview\_and\_Related\_Data\_AccessTransfer\_Issues

G. Haase

Wednesday 13<sup>th</sup> December, 2023, 14:45

<sup>&</sup>lt;sup>23</sup>https://www.packtpub.com/eu/application-development/cuda-cookbook

<sup>&</sup>lt;sup>24</sup>https://github.com/PacktPublishing/Learn-CUDA-Programming

<sup>&</sup>lt;sup>25</sup>https://static.packt-cdn.com/downloads/9781788996242\_ColorImages.pdf

<sup>&</sup>lt;sup>26</sup>https://www.researchgate.net/publication/326920953\_Unified\_Memory\_in\_CUDA\_6\_A\_Brief\_

<sup>&</sup>lt;sup>27</sup>https://developer.nvidia.com/sites/default/files/akamai/cuda/files/Misc/mygpu.pdf