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Phase 0: Understanding of LSQR method

Phase 1: Programm and test of the LSQR method with a small matrix

Phase 2: Try and modify LSQR method for a matrix A that can not fully stored in the memory

What we have done:

• Solving the LSQR problem (Yuval):

- Look up for useful libraries
- Implementing the LSQR problem in CUDA
- Solving compilation & linking issues, debugging and verifying correctness

• <u>Testing (Dominik):</u>

- Find suitable and different test matrices and vectors
- Implement the given LSMR library for any input data
- Compare results from the CUDA implementing and the LSMR library

LSQR implementation

The algorithm in the article

Algorithm LSQR

(1) (Initialize.)

$$\beta_1 u_1 = b, \quad \alpha_1 v_1 = A^T u_1, \quad w_1 = v_1, \quad x_0 = 0,$$

$$\bar{\phi}_1 = \beta_1, \quad \bar{\rho}_1 = \alpha_1.$$

- (2) For $i = 1, 2, 3, \dots$ repeat steps 3-6.
- (3) (Continue the bidiagonalization.)

(a)
$$\beta_{i+1}u_{i+1} = Av_i - \alpha_i u_i$$

(b)
$$\alpha_{i+1}v_{i+1} = A^{T}u_{i+1} - \beta_{i+1}v_{i}$$

(4) (Construct and apply next orthogonal transformation.)

(a)
$$\rho_i = (\bar{\rho}_i^2 + \beta_{i+1}^2)^{1/2}$$

(b)
$$c_i = \bar{\rho_i}/\rho_i$$

(c)
$$s_i = \beta_{i+1}/\rho_i$$

(d)
$$\theta_{i+1} = s_i \alpha_{i+1}$$

(e)
$$\bar{\rho}_{i+1} = -c_i \alpha_{i+1}$$

$$(f) \phi_i = c_i \bar{\phi_i}$$

(g)
$$\vec{\phi}_{i+1} = s_i \vec{\phi}_i$$
.

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(5) (Update x, w.)

(a)
$$x_i = x_{i-1} + (\phi_i/\rho_i)w_i$$

(b)
$$w_{i+1} = v_{i+1} - (\theta_{i+1}/\rho_i)w_i$$
.

(6) (Test for convergence.)

Pseudo-Code

```
beta = norm(b):
u = b / beta;
v = At*u:
alpha = norm(v);
v = v/alpha;
w = v:
x = 0:
phi hat = beta;
rho hat = alpha;
while (not converged) {
   //bidiagnolization
   u = A * v - alpha * u;
   beta = norm(u);
    u = u / beta;
    v = At * u - beta * v;
    alpha = norm(v);
   v = v / alpha;
   // orthogonal transformation
   rho = sqrt(rho hat^2 + beta^2);
    c = rho hat / rho;
   s = beta / rho;
    theta = s * alpha;
    rho hat = -c * alpha;
   phi = c * phi hat;
   phi hat = s * phi hat;
    //update next vectors
    x = x + (phi / rho) * w;
    w = v - (theta / rho) * w;
    residual = norm(A*x - b);
    checkForConvergnce();
```

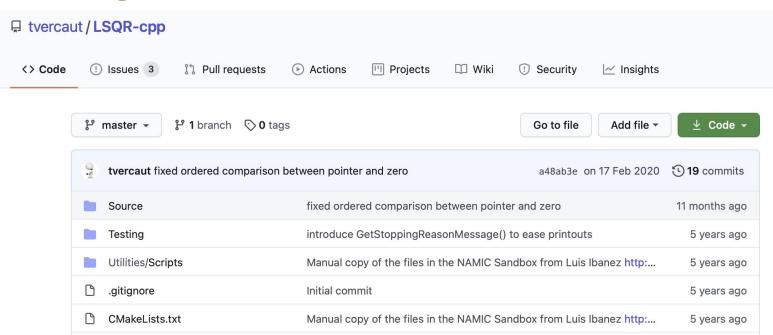
Implementation in cuBLAS

```
//Ax - b (result in tempVector)
tempDouble = -1.0;
tempDouble2 = 1.0;
status = cublasDgemv (handle, CUBL
cuBLASCheck(status,__LINE__);
status = cublasDnrm2(handle, tempV
cuBLASCheck(status,__LINE__);
improvment = prev_err-curr_err;
printf("line: %d size of error: %.
if(improvment<ebs) counter++; else
if(counter>1000) break;
prev_err = curr_err;
```

Why cuBLAS?

- Wide range of linear algebra functions (All the required Matrix & Vector operations)
- Tested and trusted library from NVIDIA
- Most of the functions have a cuBLAS equivalent (e.g. cublasDaxpy -> cusparseDaxpyi)
- Allows debugging while still working in dense format
- Requires no special data type or data structure to work with it
- Has both float and double precision function

Testing



Testing

- We are checking:
 - The results
 - Number of iteration
 - Stopping reason

- Norm of final value of residuals
- Norm of final solution

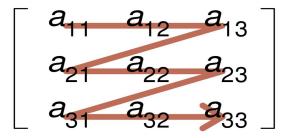
```
**std::cout << "Stopped because " << solver.GetStoppingReason() << ": " << solver.GetStoppingReasonMessage() << std::endl;
**std::cout << "Used " << solver.GetNumberOfIterationsPerformed() << " Iterations" << std::endl;
**std::cout << "Estimate of final value of norm of residuals = " << solver.GetFinalEstimateOfNormOfResiduals() << std::endl;
**std::cout << "Estimate of norm of final solution = " << solver.GetFinalEstimateOfNormOfX() << std::endl;</pre>
```

- How do we test?
 - Predefined matrices and vectors in multiple sizes

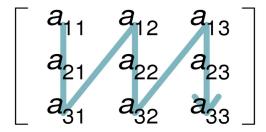
Problems

- Different format that are used in different CUDA libraries

Row-major order



Column-major order



1.1. Data layout

For maximum compatibility with existing Fortran environments, the cuBLAS library uses column-major storage, and 1-based indexing. Since C and C++ use row-major storage, applications written in these languages can not use the native array semantics for two-dimensional arrays. Instead, macros or inline functions should be defined to implement matrices on top of one-dimensional arrays. For Fortran code ported to C in mechanical fashion, one may chose to retain 1-based indexing to avoid the need to transform loops. In this case, the array index of a matrix element in row "i" and column "i" can be computed via the following macro

#define IDX2F(i,j,ld) ((((j)-1)*(ld))+((i)-1))

Problems

- Stopping conditions
- Float / double
- Linking problem
- It works for small matrix like 3x3 but not for a matrix 1000x1000
- Converging problems

Our next steps

- Write the implementation with the cuSPARSE library
- Build a parser for sparse matrices, to read from and write to files.
- Increase parallelization and memory usage of the algorithm to improve performance
- Test and verify our results



Do you have any questions?