

CS 179: Lecture 12

Recitation

cuBLAS, cuSolver, and Point Alignment

Homework 4 Released Tonight

- Will be most likely due Friday
- Due date will overlap with set 5, which will be released Wednesday as usual.
- Recommend to turn in Homework 4 by next wednesday
 - Set 4 is intentionally easy (hopefully), should give you time to prepare for midterms in other classes.

Recap

- cuBLAS is CUDA's linear algebra library!
- Good for operations between vectors, matrices, etc.
 - Feels like Matlab, Numpy, etc. E Z
 - Heavily optimized
 - Very general, can use in many applications. For example, many cublas functions subsume some of the reductions that we have been writing by hand so far such as parallelized max or parallelized sum.
- Why ever write your own kernels?
 - More control, sometimes allows for less data i/o
 - Many calls can increase overhead
 - Bad support / growing environment

Today

- Finish covering cuBlas via example
- What is cuSolver?
 - Matrix factorization
 - Parallel LU solve
- Point alignment
 - Like least squares
 - Will solve for a linear transformation that matches one set of points to another

Cublas Example

What is cuSolver

- There are primitive linear solver capabilities within cuBLAS
 - Under BLAS-like extensions
 - Mostly very primitive, not very well supported.
- cuSOLVER is entirely designed for solving linear systems
- Two big things:
 - Factorizations
 - Backsubstitution / solving factorized system
 - Great for dense linear systems

What is a linear system?

Want to solve this problem:

$$Ax = b$$

Know matrix A , know vector b , want to determine what vector x is.

Naive solution: Invert A if possible!

$$x = A^{-1}b$$

Worst solution! Bad numerical stability (same reason that $1/x$ generally unstable if x is small).

Naming Convention -- Like cuBLAS

```
cusolverDn<t><operation>
```


Also Has handle

```
cusolverStatus_t
```

```
cusolverDnCreate(cusolverDnHandle_t *handle);
```

Factorizations?

Fact:

Can factorize any matrix A into the following form:

$$P A = L U$$

L = Lower Triangular (1's on diagonal)

U = Upper triangular

P = Permutation matrix (for numerical stability)

Why useful?

Want to solve

$$Ax = b$$

$$LUx = Pb$$

=> Solve

$$Ly = Pb$$

$$Ux = y$$

Solving triangular matrices is easy!

Backsubstitution.

$$\begin{aligned}a_{1,1}x_1 + a_{1,2}x_2 + a_{1,3}x_3 + \dots + a_{1,n-1}x_{n-1} + a_{1,n}x_n &= b_1 \\a_{2,2}x_2 + a_{2,3}x_3 + \dots + a_{2,n-1}x_{n-1} + a_{2,n}x_n &= b_2 \\a_{3,3}x_3 + \dots + a_{3,n-1}x_{n-1} + a_{3,n}x_n &= b_3 \\&\vdots \\&\vdots \\&\vdots \\a_{n-1,n-1}x_{n-1} + a_{n-1,n}x_n &= b_{n-1} \\a_{n,n}x_n &= b_n\end{aligned}$$

Can solve multiple RHS simultaneously!

Multiple RHS

Furthermore, multiple RHS increases the parallelism of the application.

Can improve performance, even over fast and optimized CPU code!

<http://citeseerx.ist.psu.edu/viewdoc/summary?doi=10.1.1.2.9349>

How to solve on the hw?

cusolverDn<t>getrf()

cusolverDn<t>getrs()

Will do some of the solves simultaneously.

Point Alignment

Everything in homogeneous coordinates (bias term in ML-lingo)

Want to find matrix M of size 4 x 4

Points X1, N x 4 match to X2, N x 4

Looks like least squares, we will solve:

$$X1^T X1 \cdot M^T = X1^T X2$$

Linear system!

Homogeneous coordinates in 3D

The translation and scaling are very similar in 3D:

$$\begin{array}{l} \text{Point} \\ \vec{p} \equiv \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} \end{array} \quad \begin{array}{l} \text{Translation} \\ T = \begin{bmatrix} 1 & 0 & 0 & d_x \\ 0 & 1 & 0 & d_y \\ 0 & 0 & 1 & d_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \end{array} \quad T \cdot \vec{x} = \begin{bmatrix} 1 & 0 & 0 & d_x \\ 0 & 1 & 0 & d_y \\ 0 & 0 & 1 & d_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} x+d_x \\ y+d_y \\ z+d_z \\ 1 \end{bmatrix} = \vec{x} + \vec{d}$$

$$\begin{array}{l} \text{Scaling} \\ T = \begin{bmatrix} S_x & 0 & 0 & 0 \\ 0 & S_y & 0 & 0 \\ 0 & 0 & S_z & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \end{array} \quad T \cdot \vec{p} = \begin{bmatrix} S_x & 0 & 0 & 0 \\ 0 & S_y & 0 & 0 \\ 0 & 0 & S_z & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix} = \begin{bmatrix} S_x \cdot x \\ S_y \cdot y \\ S_z \cdot z \\ 1 \end{bmatrix} = S \cdot \vec{p}$$

Multiple RHS!

$$X_1^T X_1 \cdot M^T = X_1^T X_2$$

Minimizes distance from X_1 points transformed by M to points X_2 .

Set bottom row of M to zero with a 1 at the end.

Questions?