Solving 2D the Poisson Equation Using SBP-SAT Finite Difference Operators in Parallel

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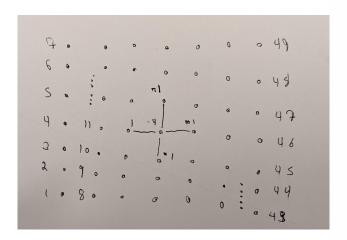
PDE Background

- Earthquake modeling with Poisson's equation
- Hooke's law and Conservation of Momentum
- Neglecting acceleration taking second time derivative to be zero results in Poisson's Equation:

$$c_1 \frac{\partial^2 u}{\partial x^2} + c_2 \frac{\partial^2 u}{\partial y^2} + F(x, y, t) = 0$$

- Goal of solving the equation is to find and equation u that satisfies the equation.
- Method of Manufactured Solutions (MMS): solution -> source functions + boundary conditions (used in solving equations)

Stencil Computations in 2D and Method of Lines



$$\begin{bmatrix} \begin{bmatrix} \vdots \\ u^1 \\ \vdots \end{bmatrix} \\ \begin{bmatrix} \vdots \\ u^2 \\ \vdots \end{bmatrix} \end{bmatrix} = \begin{bmatrix} u_1 \\ u_2 \\ u_3 \\ \vdots \\ \vdots \\ u_{n^2-2} \\ u_{n^2-1} \\ u_n \end{bmatrix}$$

$$\frac{\partial^2 u}{\partial x^2} \approx \frac{u_{i,j-1} - 2u_{i,j} + u_{i,j+1}}{\Delta x^2}$$

SBP-SAT

- Summation by Parts (SBP):
 - Would like to ensure that spatial discretization results in a "stable" system.
 - O Take energy estimate of differential equation (advection example): $\frac{\partial u}{\partial t} + a \frac{\partial u}{\partial x} = 0$

$$\frac{d}{dt} \int u^2 dx = \int \frac{\partial u^2}{\partial t} dx = \int 2u \frac{\partial u}{\partial t} dx = 2 \int u - a \frac{\partial u}{\partial t} = -a(U_r^2 - u_l^2)$$

Have discrete operators mimic continuous energy.

Need numerical quadrature, and boundary operators.

- Simultaneous approximation terms (SAT)
 - Injection destroys SBP
 - SATs minimize dependence between blocks of discretizations (good for parallelizing).
 - Can help model discontinuous physical properties.

Time and Space Complexity of Implementations

- Must solve Ax = b
- LU decomposition: Does not necessarily maintain sparsity. For big problems there could be storage constraints. Faster than CG a lot of the time. O(n^3) for factorization and O(n^2) for solve but not parallelizable (efficiently).
- CG with sparse matrices (CPU and GPU): Only need to store sparse A matrix and b resulting in better storage. Potentially slower than LU.
- CG matrix-free (CPU and GPU): Instead of storing A, just write functions that act on x, the way A would. Only storage constraint is b and intermediate arrays.

Implementations

CPU (Toby Harvey): Sparse Matrix Vector Product (SpMV, CSR), Matrix-Free

GPU (Alexandre Chen): Matrix-Free with Shared_memory, SpMV (using Toby's CPU CSR data and

SpMV_GPU() function)

Performance Analysis (Both):

CPU implementation

- Construct all operators (embarrassingly parallel for matrix verison)
- Use parallel spmv within CG kernel, otherwise call matrix free operator which takes in the vector pk produces how A acts on in Ap.

```
spmv(csr_row_ptr, csr_col_ind, csr_vals, m, n, nnz, pk, Ap);
Amv (n, pk, Ap);
```

GPU kernels

Naive kernel

- 1. 1D kernel
- 2. No shared memory
- 3. Using global index to determine boundary and interior points

GPU kernels

```
template <class T>
global void
D2x kernel(T* idata, T* odata, int Nx, int Ny, double h)
   unsigned int idx = blockIdx.x * blockDim.x + threadIdx.x;
   int N;
   N = Nx * Ny;
   if (idx < Ny)
       odata[idx] = (idata[idx] - 2*idata[idx + Ny] + idata[idx + 2*Ny]) / (h*h);
   if ((Ny \le idx) \&\& (idx < N - Ny))
       odata[idx] = (idata[idx - Ny] - 2*idata[idx] + idata[idx + Ny]) / (h*h);
   if ((idx >= N - Ny) \&\& (idx < N))
       odata[idx] = (idata[idx - 2*Ny] - 2*idata[idx - Ny] + idata[idx]) / (h*h);
   __syncthreads();
```

GPU kernels

2D shared memory kernels

- 1. Mapping: 2D real domain to 2D block and tiles.
- 2. Reading global data to static shared memory and doing calculation using shared memory
- 3. Global index with local thread index for determining boundary and interior points

2D static shared memory

```
D2x_shared_kernel(T* idata, T* odata, int Nx, int Ny, double h)
{
  const int HALO_WIDTH = 2;
  const int TILE_DIM_1 = 4; // will be set globally
  const int TILE_DIM_2 = 4;
  __shared__ double tile[TILE_DIM_1][TILE_DIM_2 + 2 * HALO_WIDTH]; // calculation in x direction
```

Mapping

Loading data from global memory into shared memory

```
// for tile itself
  if ((k < TILE DIM 1) && (l < TILE DIM 2) && (i < Ny) && (j < Nx))
      // for left halo
  if ( (k < TILE DIM 1) && (1 < HALO WIDTH) && (i < Ny) && (j >= HALO WIDTH) && (j < HALO WIDTH + Nx))
      tile[k][l] = idata[global index - HALO WIDTH * Ny]; // left-most tile doesn't need to fill left halo
  // for right halo
  if ((k < TILE DIM 1) && (l >= TILE DIM 2 - HALO WIDTH) && (l < TILE DIM 2) && (i < Ny) && ( j < Nx - HALO WIDTH))
      tile[k][1+2*HALO WIDTH] = idata[global index + HALO WIDTH*Ny]; // right-most tile doesn't need to fill right
halo
  syncthreads();
```

Calculating Finite Difference Using Data loaded to shared memory

```
// left boundary for second order
if ((k < TILE_DIM_1) && (l + HALO_WIDTH < TILE_DIM_2 + 2*HALO_WIDTH - 2) && (i < Ny) && (j == 0)) {
    odata[global_index] = (tile[k][l + HALO_WIDTH] - 2*tile[k][l+HALO_WIDTH+1] + tile[k][l+HALO_WIDTH+2]) / (h*h);
} // We need to make sure tile index and global_index satisfy certain criteria

// center
if ((k < TILE_DIM_1) && (l + HALO_WIDTH < TILE_DIM_2 + 2*HALO_WIDTH - 1) && (i < Ny) && (j >= 1) && (j < Nx - 1)) {
    odata[global_index] = (tile[k][l + HALO_WIDTH - 1] - 2*tile[k][l+HALO_WIDTH] + tile[k][l+HALO_WIDTH + 1]) / (h*h);
}

// right boundary for second order
if ((k < TILE_DIM_1) && (2 <= 1 + HALO_WIDTH) && (1 + HALO_WIDTH < TILE_DIM_2 + 2*HALO_WIDTH) && (i < Ny) && (j == Nx-1)) {
    odata[global_index] = (tile[k][l+HALO_WIDTH - 2] - 2*tile[k][l + HALO_WIDTH - 1] + tile[k][l+HALO_WIDTH]) / (h*h);
}</pre>
```

Matrix-Free Function

- 1. One GPU function that calls multiple kernels for different operators to replace spmv_gpu() in CG.
- Using vec_add_gpu() to merge all outputs. Might need to write multi_vec_add_gpu().
- 3. Ideal design:
 - a. Merging D2x and D2y kernels into one kernel with four halo regions.
 - b. Merging boundary operators into four (Dx, Dy, Hxinv, ...)
 - c. Reducing memory allocations for intermediate results
 - d. Doing boundary operations on CPU?

Current GPU Matrix-Free Function:

```
void matrix free(int Nx, int Ny, double h, double* dx, double* db, double* ...) // more variables
   unsigned TILE DIM 1 = 4;
   unsigned TILE DIM 2 = 4; // we could explore different tile shapes
   dim3 dimBlock 2d x(TILE DIM 1, TILE DIM 2, 1);
   dim3 dimGrid 2d x(Nx/TILE DIM 1 + 1, Ny/TILE DIM 2 + 1,1);
   dim3 dimBlock 2d y(TILE DIM 2, TILE DIM 1, 1);
   dim3 dimGrid 2d x(Nx/TILE DIM 1 + 1, Ny/TILE DIM 2 + 1,1);
   D2x shared kernel<double><<dimGrid 2d x, dimBlock 2d x>>>(dx, D2x out, Nx, Ny, h);
   D2y shared kernel<double><<<dimGrid 2d, dimBlock 2d>>>(dx,D2y out,Nx,Ny,h);
  vec add gpu(Nx*Ny, 1.0, D2x out, D2y out, D2 out); // Combine D2x out and D2y out;
```

Progress So far

- 1. All kernel functions implemented and tested.
- 2. Prototype for one big Matrix-free GPU function
- 3. Convergence test on CPU code ...

To Do

- 1. Finished implementation by Dec 3rd.
- 2. More test for performance and convergence for project paper.

3.