ACM/CS 114 Parallel algorithms for scientific applications

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Winter 2012

Solving a simple PDE on a uniform structured grid

Laplace equation over some domain $\Omega \in \mathbb{R}^d$, subject to Dirichlet boundary conditions

$$\nabla^2 \phi = 0 \quad \text{with} \quad \phi(\partial \Omega) = f \tag{1}$$

- let the grid be uniform: $\delta_x = \delta_y$
- ▶ in two dimensions, using first order central differences, Eq. 1 becomes

$$(\partial_{xx} + \partial_{yy}) = 0 (2)$$

▶ and translates into the following constraint among grid elements

$$= \frac{1}{4} \tag{4}$$

using a shorthand for the sum of the neighboring cells



An example

- specifically,
 - \blacktriangleright let Ω be the unit box in two dimensions
 - ightharpoonup and let ϕ satisfy the following boundary conditions

the exact solution is given by

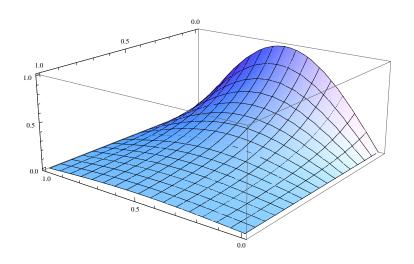
$$\phi(x,y) = e^{-\pi y} \sin(\pi x) \tag{6}$$

- ▶ we will solve this equation using the Jacobi iterative scheme:
 - make an initial guess for ϕ over a discretization of Ω
 - apply the boundary conditions
 - ▶ interpret Eq. 4 as an update step to compute the next iteration

stop when a convergence criterion is met



The solution



Implementation strategy

- grid resolution:
 - ideally determined by analyzing the boundary conditions, since discrete sampling may wash out sharp features
 - ▶ for our simple example, this can be done as part of the solver initialization
 - we will use an N × N grid and let N be user specified so we can control the problem size

$$\delta_{x} = \delta_{y} = \frac{1}{N - 1} \tag{8}$$

- data layout
 - investigate the effect of data locality by trying out various layouts
- setting up the update
 - we only need to keep track of two iterants
 - can be done in place; do you see how?
- convergence criterion
 - we will stop iterating when

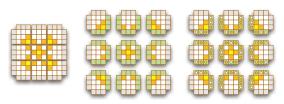
$$\max_{\Omega} (\phi_t - \phi_{t-1})) < \epsilon \tag{9}$$

and let the user specify ϵ



Parallelization

- the finest grain of work is clearly the cell update based on the value of its four nearest neighbors
- the shared memory implementation requires
 - a scheme so that threads can update cells without the need for locks
 - while maximizing locality of data access
 - even the computation of the convergence criterion can be parallelized
- ▶ with MPI
 - must partition the mesh among processes
 - each process work on its own subgrid
 - communication is required every iteration
 - parallel convergence testing involves a collective operation



Sequential implementation - user interface

```
77 // main program
  int main(int argc, char* argv[]) {
79
      // default values for our user configurable settings
      size t N = 10:
80
      double tolerance = 1.0e-6:
81
      const char* filename = "laplace.csv";
82
83
      // read the command line
84
      int command:
85
      while ((command = getopt(argc, argv, "N:e:o:")) != -1) {
86
         switch (command) {
         // get the convergence tolerance
88
         case 'e':
89
            tolerance = atof(optarg);
90
            break:
91
         // get the grid size
         case 'N':
93
            N = (size_t) atof(optarg);
95
            break;
         // get the name of the output file
96
         case 'o':
            filename = optarg;
98
99
100
```

Sequential implementation - driving the solver

```
102
      // allocate space for the solution
      Grid potential (N);
      // initialize and apply our boundary conditions
      initialize (potential) ;
106
      // call the solver
108
      laplace(potential, tolerance);
109
110
      // open a stream to hold the answer
      std::fstream output(filename, std::ios base::out);
      // build a visualizer and render the solution in our chosen format
114
      Visualizer visualizer;
      visualizer.csv(potential, output);
116
118
      // all done
      return 0:
119
120 }
```

Sequential implementation - the preamble

back up to the beginning of the file

```
#include <getopt.h>
#include <cmath>
#include <cstdlib>
#include <fstream>
#include <iostream>

// forward declarations
class Grid;
class Visualizer;

// the solver; does nothing for the time being
void initialize(Grid & grid) {};
void laplace(Grid & grid, double tolerance) {};
```

- we have separated out visualization in a different object to support different formats without disturbing the data representation
- ▶ initialize and laplace have trivial implementations for now
 - enables testing the scaffolding without worrying about the solver implementation just yet

Sequential implementation - the grid object stub

```
15 // the solution representation
16 class Grid {
    // interface: TBD
  public:
     // meta methods
20
  public:
     Grid(size_t size);
    ~Grid();
     // private data members: TBD
25
  private:
     // disabled interface
28
     // grid will own dynamic memory, so don't let the compiler screw up
30 private:
     Grid(const Grid &):
     const Grid & operator= (const Grid &);
33 };
34
35 // the grid implementation
  Grid::Grid(size_t size) {
38
39 Grid:: ~Grid() {
  }
40
```

Sequential implementation - the visualizer stub

```
97 // the visualizer class
98 class Visualizer {
     // local type aliases
  public:
     typedef std::ostream stream t;
     // interface
  public:
     void csv(const Grid & grid, stream t & stream);
106
     // meta methods
  public:
      inline Visualizer() {}
110
  1:
  // the Visualizer class implementation
  void Visualizer::csv(const Grid & grid, Visualizer::stream_t & stream) {
      return;
115 }
```

- the code now compiles and links
 - consistency check that the object collaborations are ok, for now
 - can be tested for command line option parsing