

Vaango Application Development

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and

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Contents

1	Overview of the Uintah Framework	5
1.1	Scheduler	7
1.1.1	needRecompile()	7
1.2	Tasks	7
1.3	Tasks and Scheduler Description – Programmer Interface	8
1.3.1	Simulation Component Class Description	8
1.3.2	Data Storage Concepts	9
1.4	Grid	10
1.4.1	Cells	10
1.4.2	Patches	11
1.4.3	Indexing	13
2	Examples	15
2.1	Poisson1	15
2.1.1	Description of Scheduling Functions	16
2.1.2	Description of Computational Functions	18
2.2	Poisson2	21
2.3	Burger	23
2.4	HeatEquation	25
2.5	HeatEquation Using Petsc Solver Interface	26
2.6	WaveEquation	26
2.7	AdvectionDiffusion	26
2.8	Specifying Boundary Conditions	26
2.9	Input File Specification	26
2.10	Output File Format	26

3	Implementing a New Simulation Component	27
4	Saving Simulation Data	29
5	Debugging	31
5.1	Debugger	31
5.1.1	Serial Debugging	31
5.1.2	Parallel Debugging	31
5.1.3	Running SUS	32
5.2	Memory Management	32
5.2.1	scinew	32
	Bibliography	33



1 — Overview of the Uintah Framework

The Uintah Computational Framework, i.e. **Uintah** consists of a set of software components and libraries that facilitate the solution of Partial Differential Equations (PDEs) on Structured AMR (SAMR) grids using hundreds to thousands of processors.

One of the challenges in designing a parallel, component-based multi-physics application is determining how to efficiently decompose the problem domain. Components, by definition, make local decisions. Yet parallel efficiency is only obtained through a globally optimal domain decomposition and scheduling of computational tasks. Typical techniques include allocating disjoint sets of processing resources to each component, or defining a single domain decomposition that is a compromise between the ideal load balance of multiple components. However, neither of these techniques will achieve maximum efficiency for complex multi-physics problems.

Uintah uses a non-traditional approach to achieving parallelism, employing an abstract taskgraph representation to describe computation and communication. The taskgraph is an explicit representation of the computation and communication that occur in the course of a single iteration of the simulation (typically a timestep or nonlinear solver iteration) see figure 1.1. Uintah components delegate decisions about parallelism to a scheduler component, using variable dependencies to describe communication patterns and characterizing computational workloads to facilitate a global resource optimization. The taskgraph representation has a number of advantages, including efficient fine-grained coupling of multi-physics components, flexible load balancing mechanisms and a separation of application concerns from parallelism concerns. However, it creates a challenge for scalability which we overcome by creating an implicit definition of this graph and representing it in a distributed fashion.

The primary advantage of a component-based approach is that it facilitates the separate development of simulation algorithms, models, and infrastructure. Components of the simulation can evolve independently. The component-based architecture allows pieces of the system to be implemented in a rudimentary form at first and then evolve as the technologies mature. Most importantly, Uintah allows the aspects of parallelism (schedulers, load-balancers, parallel input/output, and so forth) to evolve independently of the simulation components. Furthermore, components enable replacement of computation pieces without complex decision logic in the code itself.

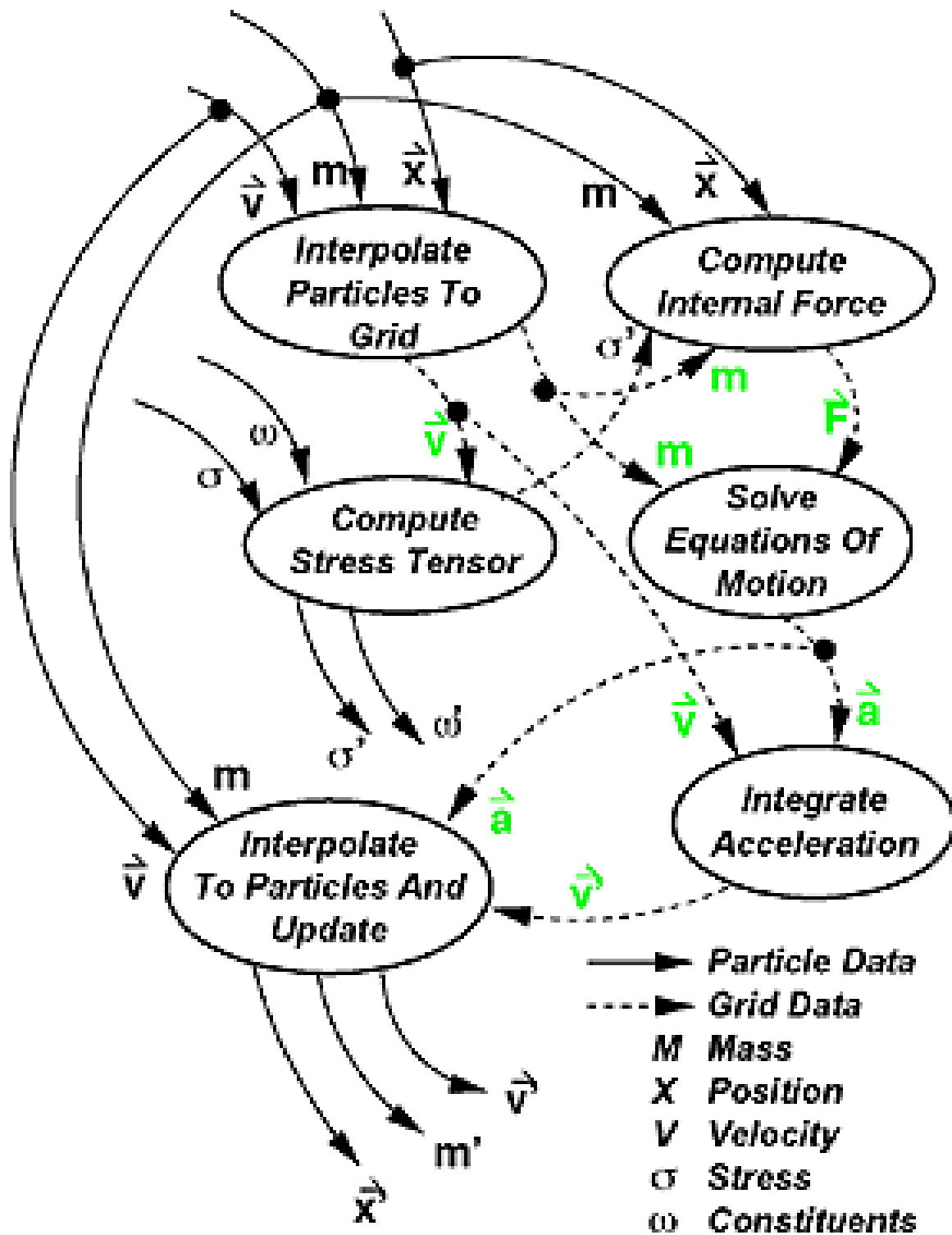


Figure 1.1: Example Task Graph

1.1 Scheduler

The Scheduler in Uintah is responsible for determining the order of tasks and ensuring that the correct inter-processor data is made available when necessary. Each software component passes a set of tasks to the scheduler. Each task is responsible for computing some subset of variables, and may require previously computed variables, possibly from different processors. The scheduler will then compile this task information into a task graph, and the task graph will contain a sorted order of tasks, along with any information necessary to perform inter-process communication via MPI. Then, when the scheduler is executed, the tasks will execute in the pre-determined order.

1.1.1 needRecompile()

Each component has a `needRecompile()` function that is called once per timestep. If, for whatever reason, a Component determines that the list of tasks it had previously scheduled is no longer valid, then the Component must return 'true' when its `needRecompile()` function is called. This will cause the scheduler to rebuild the task graph (by asking each component to re-specify tasks). Note, rebuilding the taskgraph is a relatively expensive operation, so only should be done if necessary.

1.2 Tasks

A task contains two essential components: a pointer to a function that performs the actual computations, and the data inputs and outputs, i.e. the data dependencies required by the function. When a task requests a previously computed variable from the data warehouse, the number of ghost cells are also specified. The Uintah framework uses the ghost cell information to execute inter-process communication to retrieve the necessary ghost cell data.

An example of a task description is presented showing the essential features that are commonly used by the application developer when implementing an algorithm within the Uintah framework. The task component is assigned a name and in this particular example, it is called `taskexample` and a function pointer, `&Example::taskexample`. Following the instantiation of the task itself, the dependency information is assigned to the tasks. In the following example, the task requires data from the previous timestep (`Task::OldDW`) associated with the name `variable1_label` and requires one ghost node (`Ghost::AroundNodes, 1`) level of information which will be retrieved from another processor via MPI. In addition, the task will compute two new pieces of data each associated with different variables, i.e. `variable1_label`, and `variable2_label`. Finally, the task is added to the scheduler component with specifications about what patches and materials are associated with the actual computation.

```
Task* task = scinew Task("Example::taskexample", this,
&Example::taskexample);
task->requires(Task::OldDW, variable1_label, Ghost::AroundNodes, 1);
task->computes(variable1_label);
task->computes(variable2_label);
sched->addTask(task, level->eachPatch(), sharedState_->allMaterials());
```

For more complex problems involving multiple materials and multi-physics calculations, a subset of the materials may only be used in the calculation of particular tasks. The Uintah framework allows for the independent scheduling and computation of multi-material within a multi-physics calculation.

1.3 Tasks and Scheduler Description – Programmer Interface

1.3.1 Simulation Component Class Description

Each Uintah component can be described as a C++ class that is derived from two other classes: **UintahParallelComponent** and a **SimulationInterface**. The new derived class must provide the following virtual methods: `problemSetup`, `scheduleInitialize`, `scheduleComputeStableTimestep`, and `scheduleTimeAdvance`. Here is an example of the typical *.h file that needs to be created for a new component.

```
class Example : public UintahParallelComponent, public SimulationInterface {
public:

    virtual void problemSetup(const ProblemSpecP& params,
                             const ProblemSpecP& restart_prob_spec,
                             GridP& grid, SimulationStateP&);

    virtual void scheduleInitialize(const LevelP& level, SchedulerP& sched);

    virtual void scheduleComputeStableTimestep(const LevelP& level,
                                              SchedulerP&);

    virtual void scheduleTimeAdvance(const LevelP& level, SchedulerP&);

private:
    Example(const ProcessorGroup* myworld);
    virtual ~Example();

    void initialize(const ProcessorGroup*,
                  const PatchSubset* patches,
                  const MaterialSubset* matls,
                  DataWarehouse* old_dw,
                  DataWarehouse* new_dw);

    void computeStableTimestep(const ProcessorGroup*,
                              const PatchSubset* patches,
                              const MaterialSubset* matls,
                              DataWarehouse* old_dw,
                              DataWarehouse* new_dw);

    void timeAdvance(const ProcessorGroup*,
                    const PatchSubset* patches,
                    const MaterialSubset* matls,
                    DataWarehouse* old_dw,
                    DataWarehouse* new_dw);
}
```

Each new component inherits from the classes **UintahParallelComponent** and **SimulationInterface**. The component overrides default implementations of various methods. The above methods are the essential functions that a new component must implement. Additional methods to do AMR will be described as more complex examples are presented.

The roles of each of the scheduling methods are described below. Each scheduling method, i.e. `scheduleInitialize`, `scheduleComputeStableTimestep`, and `scheduleTimeAdvance` describe

ProblemSetup

The purpose of this method is to read a problem specification which requires a minimum of information about the grid used, time information, i.e. time step size, length of time for simulation, etc, and where and what data is actually saved. Depending on the problem that is solved, the input file can be rather complex, and this method would evolve to establish any and all parameters needed to initially setup the problem.

ScheduleInitialize

The purpose of this method is to initialize the grid data with values read in from the problemSetup and to define what variables are actually computed in the TimeAdvance stage of the simulation. A task is defined which references a function pointer called `initialize`.

ScheduleComputeStableTimestep

The purpose of this method is to compute the next timestep in the simulation. A task is defined which references a function pointer called `computeStableTimestep`.

ScheduleTimeAdvance

The purpose of this method is to schedule the actual algorithmic implementation. For simple algorithms, there is only one task defined with a minimal set of data dependencies specified. However, for more complicated algorithms, the best way to schedule the algorithm is to break it down into individual tasks. Each task of the algorithm will have its own data dependencies and function pointers that reference individual computational methods.

1.3.2 Data Storage Concepts

During the course of the simulation, data is computed and stored in a data structure called the DataWarehouse. The DataWarehouse is an abstraction for storage, retrieval, and access of Uintah simulation data. The Data warehouse presents a localized view of the global data distribution.

Data Archiver

The Data Archiver is a component of the UCF that allows for reading, saving, and accessing simulation data. It presents a global shared memory abstraction to the simulation data. The component developer does not have to worry about retrieving data that has been produced on a remote processor or sending data from a local processor to another processor for use. The UCF takes care of these tasks implicitly for the component.

Two Data Warehouses

During each time step (assuming a single level problem), there are two data warehouses associated with the simulations. The “new_dw” and the “old_dw” (as they are commonly referred to in the code). The “old_dw” contains data that was generated in the previous timestep, and may not be modified. Data that is generated during the current timestep will be placed in the “new_dw”. At the end of a timestep, the current “old_dw” is deleted, and then replaced with the current “new_dw”. Then a new (empty) “new_dw” is created.

At the end of a time step, some data is automatically transferred from the current time step’s “new_dw” into the next time step’s “old_dw”.

Storing and Retrieving Variables

In order to store data to the data warehouse, or to retrieve data, several things are required. First, the task (which procedure wishes access to the data) must have registered this information with the Scheduler

during task creation time. Then, inside of the task, the following call is used to pull the data from the data warehouse:

```
SFCXVariable<double> uVelocity;
```

```
new_dw->get( uVelocity, d_lab->d_uVelocitySPBCLabel, matlIndex, patch, Ghost::AroundFaces, A
```

Similarly, to put data into the data warehouse, use this call:

```
PerPatch<CellInformationP> cellInfoP;
```

```
new_dw->put( cellInfoP, d_lab->d_cellInfoLabel, matlIndex, patch );
```

Note, you cannot put data into the "old_dw".

Variables

There are three (general) types of variables in Uintah: Particle, Grid, and Reduction. Each type of variable is discussed below. Remember, in order to interact with the Data Warehouse, each variable must have a corresponding label.

Particle variables contain information about particles.

A grid variable is a representation of data across (usually) the entire computational domain. It can be used to represent temperature, volume, velocity, etc. It is implemented using a 3D array.

A "reduction", in terms of distributed software (using, for example, MPI) is a point in which all (or some defined set of) processors all communicate with each other. It usually is an expensive operation (because it requires all processors to synchronize with each other and pass data). However, many times this operation cannot be avoided. Uintah provides a "Reduction Variable" to facilitate this operation. Common reductions operations include finding the min or max of a single number on all processors, or creating the sum of a number from every processor, and returning the result to all processors.

Variables can be stored at several different locations. Variables are defined with respect to how they relate to a single cell in the computational domain. They may be located on the faces (FC) or nodes (NC) of the cell, or in the center (CC) of the cell (see Figure 1.2).

1.4 Grid

Data (Variables) used by the UCF are stored on the 'Grid'. The Grid is comprised of one or more (in the case of AMR) Levels. Each level contains one or more patches; and each patch contains a number of cells. The simulation data itself exists in (at) each cell (and is stored in a 'variable' which is implemented as a 3D array of the data in each cell across the patch). [edit]

1.4.1 Cells

n Data in (at) each cell can be specified in several ways. Specifically the data can be Node centered (NC), Cell centered (CC), or Face centered (FC). For CC data, there is one value associated with each cell; for NC data, 4 values; and for FC data, 6 values.

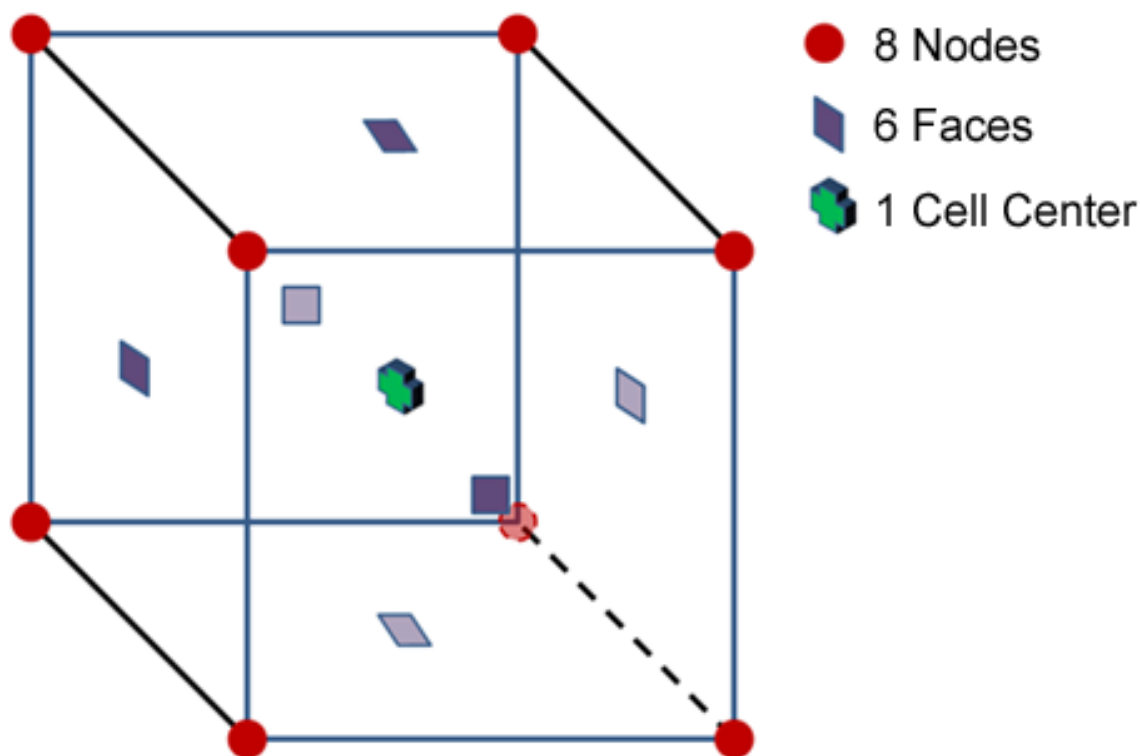


Figure 1.2: Variable locations with respect to a single cell.

1.4.2 Patches

For calculations to take place on a patch, information from bordering patches is required. This information is stored in boundary cells.

Boundary Cells

Boundary cells (Figure 1.3) represent portions of the computational domain outside the boundaries of the data assigned to a given processor. Specifically, they are almost always used as boundary cells to a patch that are necessary for computation, but are not "owned" by the patch that is currently being computed. As can be seen in Figure 1.3, patch A owns all the cells within the orange rectangle, but also has two layers of ghost cells. Portions of these cells are actually owned by patches B, C, and D. (The data found in these cells is automatically transferred (by the UCF) from processors 2, 3, and 4 to processor 1 in order for the cell data to be available for use in computations on patch A.)

Many operations will require a stencil consisting of several cells in order to perform calculations at a given cell. However, at the border of a patch, there are no cells belonging to that patch that contain the required information - that information is "owned" by another processor. In this situation, data that belongs to another patch must be accessed. This is the purpose of ghost cells. The Data Warehouse takes care of moving the required ghost data from the "owner" processor to the neighbor processor so that the individual task can assume the required data is available.

In summary, ghost cells exist between patches on the same level. They are cells that are owned by the neighboring patch but are required for computation due to the stencil width.

Extra cells exist at the edge of the computational domain and are used for boundary conditions. Extra cells exist on the boundaries of the level where no neighboring patches exist. This can either be the edge

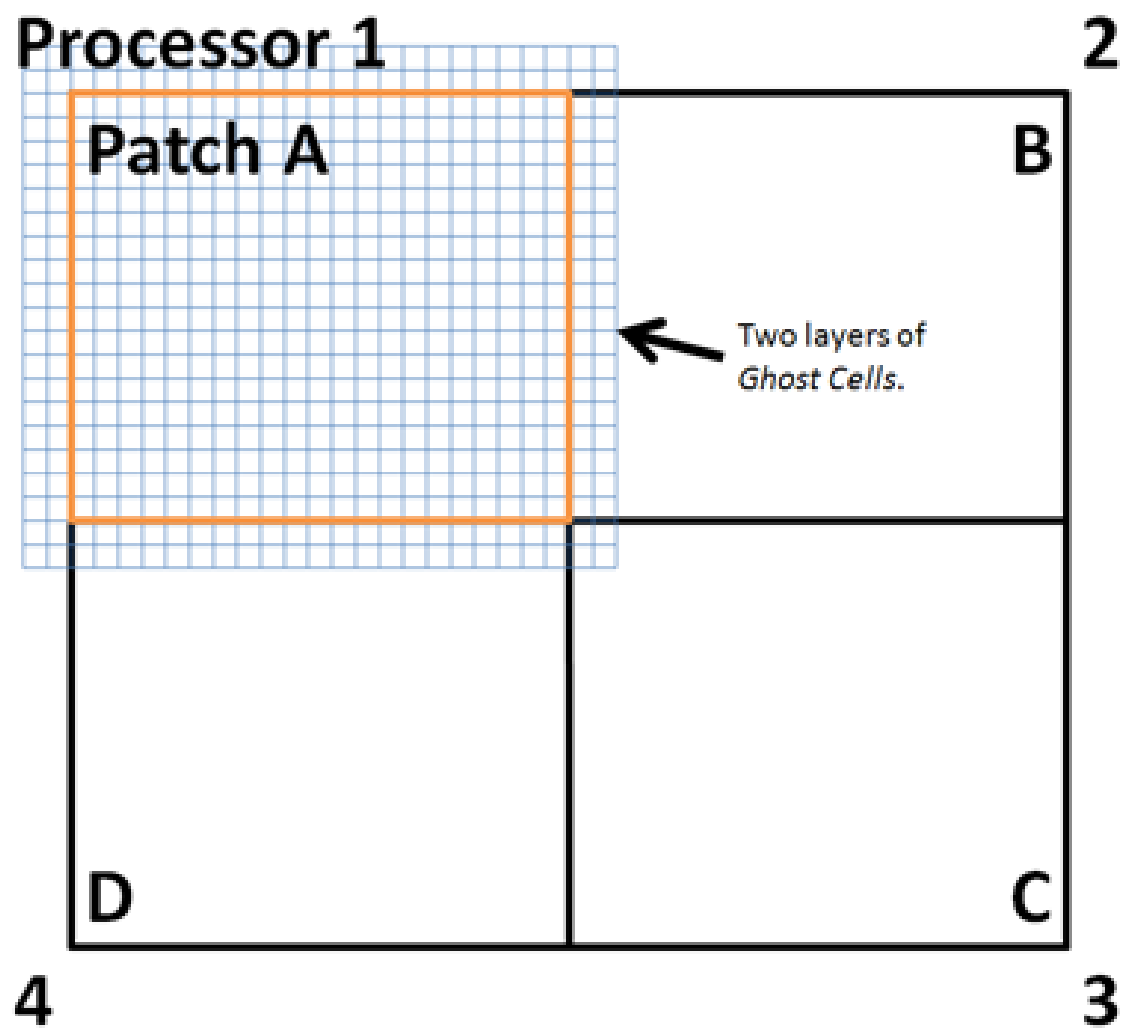


Figure 1.3: Schematic of ghost cell locations laid out with respect to patches.

of a domain or on a coarse-fine interface.

1.4.3 Indexing

Untah uses a straightforward indexing scheme. Each cell on a given level is uniquely identified by an x,y,z coordinate (stored in an `IntVector`). An `IntVector` is a vector of 3 integers that represent an X,Y,Z coordinate. See `Core/Grid/Level.cc/h` for more information. The following pseudo code shows how indices across levels are mapped to each other.

```
IntVector
Level::mapCellToCoarser(const IntVector& idx) const
{
    IntVector ratio = idx/d_refinementRatio;

    return ratio;
}

IntVector
Level::mapCellToFiner(const IntVector& idx) const
{
    IntVector r_ratio = grid->getLevel(d_index+1)->d_refinementRatio;
    IntVector fineCell = idx*r_ratio;

    return fineCell;
}
```




2 — Examples

This chapter will describe a set of example problems showing various stages of algorithm complexity and how the Uintah framework is used to solve the discretized form of the solutions. Emphasis will not be on the most efficient or fast algorithms, but instead will demonstrate straightforward implementations of well known algorithms within the Uintah Framework. Several examples will be given that show an increasing level of complexity that will serve as a guide to others interested in implementing structured AMR algorithms for PDEs.

All examples described are found in the directory *SCIRun/src/Packages/Uintah/CCA/Components/Examples*

2.1 Poisson1

Poisson1 solves Poisson's equation on a grid using Jacobi iteration. Since this is not a time dependent problem and Uintah is fundamentally designed for time dependent problems, each Jacobi iteration is considered to be a timestep. The timestep specified and computed is a fixed value obtained from the input file and has no bearing on the actual computation.

The following equation is discretized and solved using an iterative method. Each timestep is one iteration. At the end of the timestep, the residual is computed showing the convergence of the solution and the next iteration is computed until.

The following shows a simplified form of the Poisson1 of the .h and .cc files found in the Examples directory. The argument list for some of the methods are eliminated for readability purposes. Please refer to the actual source for a complete description of the arguments required for each method.

```
class Poisson1 : public UintahParallelComponent, public SimulationInterface {
public:
    Poisson1(const ProcessorGroup* myworld);
    virtual ~Poisson1();

    virtual void problemSetup(const ProblemSpecP& params,
                             const ProblemSpecP& restart_prob_spec,
                             GridP& grid, SimulationStateP&);
    virtual void scheduleInitialize(const LevelP& level, SchedulerP& sched);

    virtual void scheduleComputeStableTimestep(const LevelP& level, SchedulerP&)
        ;
```

```

virtual void scheduleTimeAdvance(const LevelP& level, SchedulerP&);

private:
void initialize(const ProcessorGroup*, const PatchSubset* patches,
const MaterialSubset* matls, DataWarehouse* old_dw,
DataWarehouse* new_dw);

void computeStableTimestep(const ProcessorGroup*, const PatchSubset* patches
,
const MaterialSubset* matls, DataWarehouse* old_dw,
DataWarehouse* new_dw);

void timeAdvance(const ProcessorGroup, const PatchSubset* patches,
const MaterialSubset* matls, DataWarehouse* old_dw,
DataWarehouse* new_dw);

SimulationStateP sharedState_;
double delt_;
const VarLabel* phi_label;
const VarLabel* residual_label;
SimpleMaterial* mymat_;

Poisson1(const Poisson1&);
Poisson1& operator=(const Poisson1&);
};

```

The private methods and data shown are the functions that are function pointers referred to in the task descriptions. The `VarLabel` data type stores the names of the various data that can be referenced uniquely by the data warehouse. The `SimulationStateP` data type is essentially a global variable that stores information about the materials that are needed by other internal Uintah framework components. `SimpleMaterial` is a data type that refers to the material properties.

Within each schedule function, i.e. `sheduleInitialize`, `scheduleComputeStableTimestep`, and `scheduleTimeAdvance`, a task is specified that has a function pointer associated with it. The function pointers point to the actual implementation of the specific task and have a different argument list than the associated schedule method.

The typical task implementation, i.e. `timeAdvance()` contains the following arguments: `ProcessorGroup`, `PatchSubset`, `MaterialSubset`, and two `DataWarehouse` objects. The purpose of the `ProcessorGroup` is to hold various MPI information such as the `MPI_Communicator`, the rank of the process and the number of processes that are actually being used.

2.1.1 Description of Scheduling Functions

The actual implementation with descriptions are presented following the code snippets.

```

Poisson1::Poisson1(const ProcessorGroup* myworld)
: UintahParallelComponent(myworld)
{
    phi_label = VarLabel::create("phi",
NCVariable<double>::getTypeDescription());
    residual_label = VarLabel::create("residual",
sum_vartype::getTypeDescription());
}

Poisson1::~~Poisson1()
{
    VarLabel::destroy(phi_label);
    VarLabel::destroy(residual_label);
}

```



```
}
```

Typical constructor and destructor for simple examples where the data label names (phi and residual) are created for data warehouse storage and retrieval.

```
void Poisson1::problemSetup(const ProblemSpecP& params,
    const ProblemSpecP& restart_prob_spec,
    GridP& /*grid*/,
    SimulationStateP& sharedState)
{
    sharedState_ = sharedState;
    ProblemSpecP poisson = params->findBlock("Poisson");

    poisson->require("delt", delt_);

    mymat_ = scinew SimpleMaterial();

    sharedState->registerSimpleMaterial(mymat_);
}
```

The problemSetup is based in a xml description of the input file. The input file is parsed and the delt tag is set. The sharedState is assigned and is used to register a material and store it for later use by the Uintah internals.

```
void Poisson1::scheduleInitialize(const LevelP& level,
    SchedulerP& sched)
{
    Task* task = scinew Task("Poisson1::initialize",
        this, &Poisson1::initialize);

    task->computes(phi_label);
    task->computes(residual_label);
    sched->addTask(task, level->eachPatch(), sharedState->allMaterials());
}
```

A task is defined which contains a name and a function pointer, i.e. initialize which is described later in Poisson1.cc. The task defines two variables that are computed in the initialize function, phi and residual. The task is then added to the scheduler. This task is only computed once at the beginning of the simulation.

```
void Poisson1::scheduleComputeStableTimestep(const LevelP& level,
    SchedulerP& sched)
{
    Task* task = scinew Task("Poisson1::computeStableTimestep",
        this, &Poisson1::computeStableTimestep);

    task->requires(Task::NewDW, residual_label);
    task->computes(sharedState->get_delt_label());
    sched->addTask(task, level->eachPatch(), sharedState->allMaterials());
}
```

A task is defined for the computing the stable timestep and uses the function pointer, computeStableTimestep defined later in Poisson1.cc. This requires data from the New DataWarehouse, and the next timestep size is computed and stored.

```
void
Poisson1::scheduleTimeAdvance( const LevelP& level,
    SchedulerP& sched)
{
    Task* task = scinew Task("Poisson1::timeAdvance",
        this, &Poisson1::timeAdvance);

    task->requires(Task::OldDW, phi_label, Ghost::AroundNodes, 1);
```

```

    task->computes(phi_label);
    task->computes(residual_label);
    sched->addTask(task, level->eachPatch(), sharedState_->allMaterials());
}

```

The `timeAdvance` function is the main function that describes the computational algorithm. For simple examples, the entire algorithm is usually defined by one task with a small set of data dependencies. However, for more complicated algorithms, it is best to break the algorithm down into a set of tasks with each task describing its own set of data dependencies.

For this example, a single task is described and the `timeAdvance` function pointer is specified. Data from the previous timestep (`OldDW`) is required for the current timestep. For a simple seven (7) point stencil, only one level of ghost cells is required. The algorithm is set up for nodal values, the ghost cells are specified by the `Ghost::AroundNodes` syntax. The task computes both the new data values for `phi` and a residual.

2.1.2 Description of Computational Functions

```

void Poisson1::computeStableTimestep(const ProcessorGroup* pg,
const PatchSubset* /*patches*/,
const MaterialSubset* /*matls*/,
DataWarehouse*,
DataWarehouse* new_dw)
{
    if(pg->myrank() == 0){
        sum_vartype residual;
        new_dw->get(residual, residual_label);
        cerr << "Residual=" << residual << '\n';
    }
    new_dw->put(delt_vartype(delt_), sharedState_->get_delt_label());
}

```

In this particular example, no timestep is actually computed, instead the original timestep specified in the input file is used and stored in the data warehouse (`new_dw->put(delt_vartype(delt_), sharedState_->get_delt_label())`). The residual computed in the main `timeAdvance` function is retrieved from the data warehouse and printed out to standard error for only the processor with a rank of 0.

```

void Poisson1::initialize(const ProcessorGroup*,
const PatchSubset* patches,
const MaterialSubset* matls,
DataWarehouse* /*old_dw*/, DataWarehouse* new_dw)
{
    int matl = 0;
    for(int p=0;p<patches->size();p++){
        const Patch* patch = patches->get(p);

```

The node centered variable (`NCVariable<double> phi`) has space reserved in the `DataWarehouse` for the given patch and the given material (`int matl = 0`). The `phi` variable is initialized to 0 for every grid node on the patch.

```

    NCVariable<double> phi;
    new_dw->allocateAndPut(phi, phi_label, matl, patch);
    phi.initialize(0.);

```

The boundary faces on the `xminus` face of the computational domain are specified and set to a value of 1. All other boundary values are set to a value of 0 as well as the internal nodes via the `phi.initialize(0.)` construct. Uintah provides helper functions for determining which nodes are on the boundaries. In addition, there are convenient looping constructs such as `NodeIterator` that alleviate the need to specify triply nested loops for visiting each node in the domain.

```

    if(patch->getBCTYPE(Patch::xminus) != Patch::Neighbor){
        IntVector l,h;
        patch->getFaceNodes(Patch::xminus, 0, l, h);

        for(NodeIterator iter(l,h); !iter.done(); iter++){
            phi[*iter]=1;
        }
    }

    new_dw->put(sum_vartype(-1), residual_label);
}
}

```

The initial residual value of -1 is stored at the beginning of the simulation.

The main computational algorithm is defined in the timeAdvance function. The overall algorithm is based on a simple Jacobi iteration step.

```

void Poisson1::timeAdvance(const ProcessorGroup*,
const PatchSubset* patches,
const MaterialSubset* matls,
DataWarehouse* old_dw,
DataWarehouse* new_dw)
{
    int matl = 0;
    for(int p=0;p<patches->size();p++){
        const Patch* patch = patches->get(p);
        constNCVariable<double> phi;
    }
}

```

Data from the previous timestep is retrieved from the data warehouse and copied to the current timestep's phi variable (newphi).

```

old_dw->get(phi, phi_label, matl, patch, Ghost::AroundNodes, 1);
NCVariable<double> newphi;

new_dw->allocateAndPut(newphi, phi_label, matl, patch);
newphi.copyPatch(phi, newphi.getLowIndex(), newphi.getHighIndex());

```

The indices for the patch are obtained and altered depending on whether or not the patch's internal boundaries are on the coincident with the grid domain. If the patch boundaries are the same as the grid domain, the boundary values are not overwritten since the lower and upper indices are modified to only specify internal nodal grid points.

```

double residual=0;
IntVector l = patch->getNodeLowIndex__New();
IntVector h = patch->getNodeHighIndex__New();

l += IntVector(patch->getBCTYPE(Patch::xminus) == Patch::Neighbor?0:1,
patch->getBCTYPE(Patch::yminus) == Patch::Neighbor?0:1,
patch->getBCTYPE(Patch::zminus) == Patch::Neighbor?0:1);
h -= IntVector(patch->getBCTYPE(Patch::xplus) == Patch::Neighbor?0:1,
patch->getBCTYPE(Patch::yplus) == Patch::Neighbor?0:1,
patch->getBCTYPE(Patch::zplus) == Patch::Neighbor?0:1);

```

The Jacobi iteration step is applied at each internal grid node. The residual is computed based on the old and new values and stored as a reduction variable (sum_vartype) in the data warehouse.

```

// -----
// Stencil
for(NodeIterator iter(l, h);!iter.done(); iter++){
    IntVector n = *iter;
}

```

```

        newphi[n]=(1./6)*(
        phi[n+IntVector(1,0,0)] + phi[n+IntVector(-1,0,0)] +
        phi[n+IntVector(0,1,0)] + phi[n+IntVector(0,-1,0)] +
        phi[n+IntVector(0,0,1)] + phi[n+IntVector(0,0,-1)]);

        double diff = newphi[n] - phi[n];
        residual += diff * diff;
    }
    new_dw->put(sum_vartype(residual), residual_label);
}
}

```

The input file that is used to run this example is given below and is given in *SCIRun/src/Packages/Uintah/StandAlone/inputs/Examples/poisson1.ups*. Relevant sections of the input file that can be modified are found in the <Time> section, and the <Grid> section, specifically, the number of patches and the grid resolution.

```

<Uintah_specification>

<Meta>
<title>Poisson1 test</title>
</Meta>

<SimulationComponent>
<type> poisson1 </type>
</SimulationComponent>
<!-- ----- -->
<Time>
<maxTime>      1.0      </maxTime>
<initTime>     0.0      </initTime>
<delt_min>     0.00001  </delt_min>
<delt_max>     1        </delt_max>
<max_Timesteps> 100     </max_Timesteps>
<timestep_multiplier> 1 </timestep_multiplier>
</Time>

<!-- ----- -->
<DataArchiver>
<filebase>poisson.uda</filebase>
<outputTimestepInterval>1</outputTimestepInterval>
<save label = "phi"/>
<save label = "residual"/>
<checkpoint cycle = "2" timestepInterval = "1"/>
</DataArchiver>

<!-- ----- -->
<Poisson>
<delt>.01</delt>
<maxresidual>.01</maxresidual>
</Poisson>

<!-- ----- -->
<Grid>
<Level>
<Box label = "1">
<lower>      [0,0,0]      </lower>
<upper>      [1.0,1.0,1.0] </upper>
<resolution> [50,50,50]   </resolution>
<patches>    [2,1,1]      </patches>
</Box>

```

```

</Level>
</Grid>

</Uintah_specification>

```

Running the Poisson1 Example

To run the poisson1.ups example,

```
cd ~/SCIRun/dbg/Packages/Uintah/StandAlone/
```

create a symbolic link to the inputs directory:

```
ln -s ~/SCIRun/src/Packages/Uintah/StandAlone/inputs
```

For a single processor run type the following:

```
sus inputs/Examples/poisson1.ups
```

For a two processor run, type the following:

```
mpirun -np 2 sus -mpi inputs/Examples/poisson1.ups
```

Changing the number of patches in the poisson1.ups and the resolution, enables you to run a more refined problem on more processors.

ADVICE: For non-AMR problems, it is advised to have at least the same number of patches as processors. You can always have more patches than processors, but you cannot have fewer patches than processors.

2.2 Poisson2

The next example also solves the Poisson's equation but instead of iterating in time, the subscheduler feature iterates within a given timestep, thus solving the problem in one timestep. The use of the subscheduler is important for implementing algorithms which solve non-linear problems which require iterating on a solution for each timestep.

The majority of the schedule and computational functions are similar to the *Poisson1* example and are not repeated. Only the revised code is presented with explanations about the new features of Uintah.

```

void Poisson2::scheduleTimeAdvance( const LevelP& level, SchedulerP& sched)
{
    Task* task = scinew Task("timeAdvance",
        this, &Poisson2::timeAdvance,
        level, sched.get_rep());
    task->hasSubScheduler();
    task->requires(Task::OldDW, phi_label, Ghost::AroundNodes, 1);
    task->computes(phi_label);
    LoadBalancer* lb = sched->getLoadBalancer();
    const PatchSet* perproc_patches = lb->getPerProcessorPatchSet(level);
    sched->addTask(task, perproc_patches, sharedState_->allMaterials());
}

```

Within this function, the task is specified with two additional arguments, the level and the scheduler `sched.get_rep()`. The task must also set the flag that a subscheduler will be used within the scheduling of the various tasks. Similar code to the **Poisson1** example is used to specify what data is required and computed during the actual task execution. In addition, a loadbalancer component is required to query the patch distribution for each level of the grid. The task is then added to the top level scheduler with the requisite information, i.e. patches and materials.

The actual implementation of the `timeAdvance` function is also different from the **Poisson1** example. The code is specified below with text explaining the use of the subscheduler. The new feature of the subscheduler shows the creation of a the `iterate` task within the subscheduler. This task will perform the actual Jacobi iteration for a given timestep.

```
void Poisson2::timeAdvance(const ProcessorGroup* pg,
    const PatchSubset* patches,
    const MaterialSubset* matls,
    DataWarehouse* old_dw, DataWarehouse* new_dw,
    LevelP level, Scheduler* sched)
{
```

The subscheduler is instantiated and initialized.

```
SchedulerP subsched = sched->createSubScheduler();
subsched->initialize();
GridP grid = level->getGrid();
```

An `iterate` task is created and added to the subscheduler. The typical computes and requires are specified for a 7 point stencil used in Jacobi iteration scheme with one layer of ghost cells. The new task is added to the subscheduler. A residual variable is only computed within the subscheduler and not passed back to the main scheduler. This is in contrast to the `phi` variable which was specified in `scheduleTimeAdvance` in the computes, as well as being specified in the computes for the subscheduler. Any variables that are only computed and used in an iterative step of an algorithm do not need to be added to the dependency specification for the top level task.

```
// Create the tasks
Task* task = scinew Task("iterate",
    this, &Poisson2::iterate);
task->requires(Task::OldDW, phi_label, Ghost::AroundNodes, 1);
task->computes(phi_label);
task->computes(residual_label);
subsched->addTask(task, level->eachPatch(), sharedState_->allMaterials());
```

The subscheduler has its own data warehouse that is separate from the top level's scheduler's data warehouse. This data warehouse must be initialized and any data from the top level's data warehouse must be passed to the subscheduler's version. This data resides in the data warehouse position `NewDW`.

```
// Compile the scheduler
subsched->advanceDataWarehouse(grid);
subsched->compile();

int count = 0;
double residual;
subsched->get_dw(1)->transferFrom(old_dw, phi_label, patches, matls);
```

Within each iteration, the following must occur for the subscheduler: the data warehouse's new data must be moved to the `OldDW` position, since any new values will be stored in `NewDW` and the old values cannot be overwritten. The `OldDW` is referred to in the subscheduler via the `subsched->get_dw(0)` and the `NewDW` is referred to in the subscheduler via `subsched->get_dw(1)`. Once the iteration is deemed to have met the tolerance, the data from the subscheduler is transferred to the scheduler's data warehouse.

```
// Iterate
do {
    subsched->advanceDataWarehouse(grid);
    subsched->get_dw(0)->setScrubbing(DataWarehouse::ScrubComplete);
    subsched->get_dw(1)->setScrubbing(DataWarehouse::ScrubNonPermanent);
    subsched->execute();

    sum_vartype residual_var;
    subsched->get_dw(1)->get(residual_var, residual_label);
```

```

    residual = residual_var;

    if(pg->myrank() == 0)
        cerr << "Iteration " << count++ << ", residual=" << residual << '\n';
    } while(residual > maxresidual_);

    new_dw->transferFrom(subsched->get_dw(1), phi_label, patches, matls);
}

```

The iteration cycle is identical to **Poisson1**'s timeAdvance algorithm using Jacobi iteration with a 7 point stencil. Refer to the discussion about the algorithm implementation in the **Poisson1** description.

```

void Poisson2::iterate(const ProcessorGroup*,
const PatchSubset* patches,
const MaterialSubset* matls,
DataWarehouse* old_dw, DataWarehouse* new_dw)
{
    for(int p=0;p<patches->size();p++){
        const Patch* patch = patches->get(p);
        for(int m = 0;m<matls->size();m++){
            int matl = matls->get(m);
            constNCVariable<double> phi;
            old_dw->get(phi, phi_label, matl, patch, Ghost::AroundNodes, 1);
            NCVariable<double> newphi;
            new_dw->allocateAndPut(newphi, phi_label, matl, patch);
            newphi.copyPatch(phi, newphi.getLow(), newphi.getHigh());
            double residual=0;
            IntVector l = patch->getNodeLowIndex__New();
            IntVector h = patch->getNodeHighIndex__New();
            l += IntVector(patch->getBCType(Patch::xminus) == Patch::Neighbor?0:1,
            patch->getBCType(Patch::yminus) == Patch::Neighbor?0:1,
            patch->getBCType(Patch::zminus) == Patch::Neighbor?0:1);
            h -= IntVector(patch->getBCType(Patch::xplus) == Patch::Neighbor?0:1,
            patch->getBCType(Patch::yplus) == Patch::Neighbor?0:1,
            patch->getBCType(Patch::zplus) == Patch::Neighbor?0:1);
            for(NodeIterator iter(l, h);!iter.done(); iter++){
                newphi[*iter]=(1./6)*(
                phi[*iter+IntVector(1,0,0)]+phi[*iter+IntVector(-1,0,0)]+
                phi[*iter+IntVector(0,1,0)]+phi[*iter+IntVector(0,-1,0)]+
                phi[*iter+IntVector(0,0,1)]+phi[*iter+IntVector(0,0,-1)]);
                double diff = newphi[*iter]-phi[*iter];
                residual += diff*diff;
            }
            new_dw->put(sum_vartype(residual), residual_label);
        }
    }
}

```

The input file `/SCIRun/src/Packages/Uintah/StandAlone/inputs/Examples/poisson2.ups` is very similar to the `poisson1.ups` file shown above. The only additional tag that is used is the `<maxresidual>` specifying the tolerance within the iteration performed in the subscheduler.

To run the `poisson2` input file execute the following in the `/SCIRun/dbg/Packages/Uintah/StandAlone` directory:

```
sus inputs/Examples/poisson2.ups
```

2.3 Burger

In this example, the inviscid Burger's equation is solved in three dimensions:

$$\frac{du}{dt} = -u \frac{du}{dx}$$

with the following initial conditions:

$$u = \sin(\pi x) + \sin(2\pi y) + \sin(3\pi z)$$

using Euler's method to advance in time. The majority of the code is very similar to the Poisson1 example code with the differences shown below.

The initialization of the grid values for the unknown variable, u , is done at every grid node using the NodeIterator construct. The x,y,z values for a given grid node is determined using the function, `patch->getNodePosition` where n is the nodal index in i,j,k space.

```
void Burger::initialize(const ProcessorGroup*,
    const PatchSubset* patches,
    const MaterialSubset* matls,
    DataWarehouse*,
    DataWarehouse* new_dw)
{
    int matl = 0;
    for(int p=0;p<patches->size();p++){
        const Patch* patch = patches->get(p);

        NCVariable<double> u;
        new_dw->allocateAndPut(u, u_label, matl, patch);

        //Initialize
        // u = sin( pi*x ) + sin( pi*2*y ) + sin(pi*3z )
        IntVector l = patch->getNodeLowIndex__New();
        IntVector h = patch->getNodeHighIndex__New();

        for( NodeIterator iter=patch->getNodeIterator__New(); !iter.done(); iter
            ++ ){
            IntVector n = *iter;
            Point p = patch->nodePosition(n);
            u[n] = sin( p.x() * 3.14159265358 ) + sin( p.y() * 2*3.14159265358 ) +
                sin( p.z() * 3*3.14159265358 );
        }
    }
}
```

The timeAdvance function is quite similar to the Poisson1's timeAdvance routine. The relevant differences are only shown.

```
void Burger::timeAdvance(const ProcessorGroup*,
    const PatchSubset* patches,
    const MaterialSubset* matls,
    DataWarehouse* old_dw,
    DataWarehouse* new_dw)
{
    int matl = 0;
    //Loop for all patches on this processor
    for(int p=0;p<patches->size();p++){
        const Patch* patch = patches->get(p);

        . . . . .
```

The grid spacing and timestep values are stored.

```
// dt, dx
Vector dx = patch->getLevel()->dCell();
delt_vartype dt;
old_dw->get(dt, sharedState_->get_delt_label());

. . . . .
```


Refer to the description in **Poisson1** about the specification of the NodeIterator limits. The Euler algorithm is applied to solve the ordinary differential equation in time.

```
//Iterate through all the nodes
for(NodeIterator iter(1, h); !iter.done(); iter++){
    IntVector n = *iter;
    double dudx = (u[n+IntVector(1,0,0)] - u[n-IntVector(1,0,0)]) / (2.0 *
        dx.x());
    double dudy = (u[n+IntVector(0,1,0)] - u[n-IntVector(0,1,0)]) / (2.0 *
        dx.y());
    double dudz = (u[n+IntVector(0,0,1)] - u[n-IntVector(0,0,1)]) / (2.0 *
        dx.z());
    double du = - u[n] * dt * (dudx + dudy + dudz);
    new_u[n] = u[n] + du;
}
```

Zero flux Neumann boundary conditions are applied to the node points on each of the grid faces.

```
//-----
// Boundary conditions: Neumann
// Iterate over the faces encompassing the domain
vector<Patch::FaceType>::const_iterator iter;
vector<Patch::FaceType> bf;
patch->getBoundaryFaces(bf);
for (iter = bf.begin(); iter != bf.end(); ++iter){
    Patch::FaceType face = *iter;

    IntVector axes = patch->faceAxes(face);
    int P_dir = axes[0]; // find the principal dir of that face

    IntVector offset(0,0,0);
    if (face == Patch::xminus || face == Patch::yminus || face == Patch::
        zminus){
        offset[P_dir] += 1;
    }
    if (face == Patch::xplus || face == Patch::yplus || face == Patch::
        zplus){
        offset[P_dir] -= 1;
    }

    Patch::FaceIteratorType FN = Patch::FaceNodes;
    for (CellIterator iter = patch->getFaceIterator__New(face, FN); !iter.
        done(); iter++){
        IntVector n = *iter;
        new_u[n] = new_u[n + offset];
    }
}
}
```

The input file for the Burger (*burger.ups*) problem is very similar to the *poisson1.ups* with the addition, that the timestep increment used in the timeAdvance is quite small, 1.e-4 for stability reasons.

To run the Burger input file execute the following in the */SCIRun/dbg/Packages/Uintah/StandAlone* directory:

```
sus inputs/Examples/burger.ups
```

2.4 HeatEquation

In this example, the simple heat equation is solved in three dimensions building upon the techniques presented earlier to implement an explicit and implicit formulation. For the implicit formulation, two

solution techniques will be demonstrated, one using the subscheduler to perform Jacobi iteration and the second will be to demonstrate using the Petsc interface for the solution of the linear equations. In addition, this example will demonstrate how material properties can be specified within the input file and used in the simulation.

$$dT/dt = k \, d^2 T/dx^2$$

Only the relevant sections of the HeatEquation.cc are shown that differ from the Poisson1/Poisson2 source files that highlight how timestepping with the subscheduler is achieved.

```
void HeatEquation::scheduleTimeAdvance( const LevelP& level, SchedulerP&
    sched)
{
    Task* task = scinew Task("timeAdvance",
        this, &Poisson2::timeAdvance,
        level, sched.get_rep());
    task->hasSubScheduler();
    task->requires(Task::OldDW, temperature_label, Ghost::AroundNodes, 1);
    task->computes(temperature_label);
    LoadBalancer* lb = sched->getLoadBalancer();
    const PatchSet* perproc_patches = lb->getPerProcessorPatchSet(level);
    sched->addTask(task, perproc_patches, sharedState_->allMaterials());
}
```

2.5 HeatEquation Using Petsc Solver Interface

In this example, heat equation will be solved using the Petsc solver interface to solve the tri-diagonal system of equations arising from the spatial component

2.6 WaveEquation

2.7 AdvectionDiffusion

2.8 Specifying Boundary Conditions

2.9 Input File Specification

The Uintah framework uses xml input files to specify the various parameters required in any simulation component. The application developer is free to use any tags to specify the data needed by the simulation. The essential tags that are required by Uintah include the following:

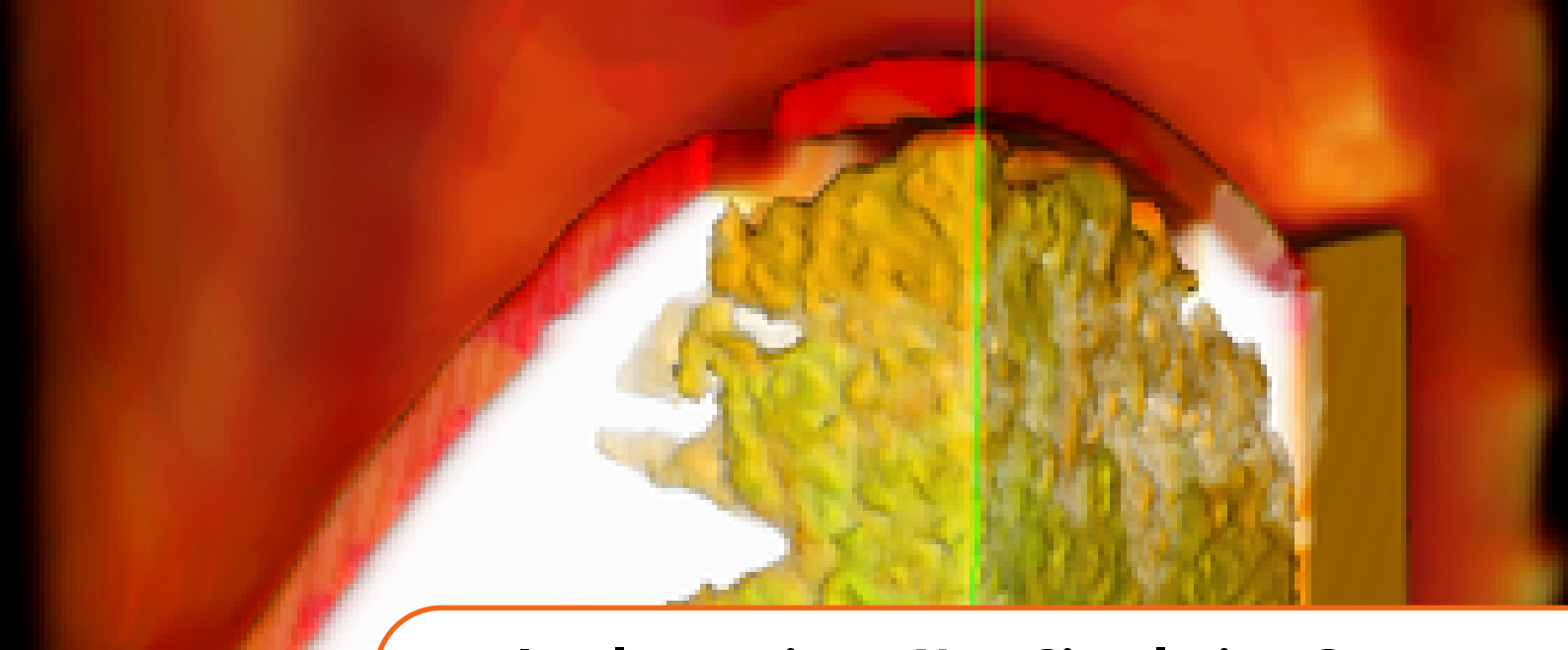
```
<Uintah_specification>

<Time>

<DataArchiver>

<Grid>
```

2.10 Output File Format



3 — Implementing a New Simulation Component

A driver program must be developed that will then be compiled and linked with the Uintah framework libraries.



4 — Saving Simulation Data

The UCF automatically saves data as specified in the .ups file. This works for all data that is stored in the Data Warehouse. However, there are times when a component will need to save its own data. (It is preferable that the Data Warehouse be updated to manage this Component data, but sometimes this is not expedient...) In these cases, here is an outline on how to save that data:

- Create a function that will save your data:

```
void
Component::saveMyData( const PatchSubset * patches )
{
    ...
}
```

- From the last `ijTaski` in your algorithm, call the function:

```
saveMyData( patches );
```

- In the function, you need to do several things:
Make sure that it is an output timestep. (Your component must have saved a pointer to the Data Archiver. Most components already do this... if not, you can do it in `Component::problemSetup()`.)

```
if( dataArchiver->isOutputTimestep() ) {
    // Dump your data... (see below)
}
```

- You need to create a directory to put the data in:

```
const int      & timestep = d_sharedState->getCurrentTopLevelTimeStep();
ostringstream  data_dir;

// This gives you ".../simulation.uda.###/t####/mydata/"
data_dir << dataArchiver()->getOutputLocation() << "/t" << setw(5) <<
    setfill('0') << timestep
<< "/mydata";

int result = MKDIR( data_dir.c_str(), 0777 ); // Make sure you #
include <Core/OS/Dir.h>

if( result != 0 ) {
    ostringstream error;
    error << "Component::saveMyData(): couldn't create directory '" +
        data_dir.str() + "' (" << result << ").";
    throw InternalError( error.str(), __FILE__, __LINE__ );
}
```

```
}

```

- Loop over the patches and save your data... (Note, depending on how much data you have, you need to determine whether you want to write the files as a binary file, or as an ascii file. Ascii files are easier for a human to look at for errors, but take a lot more time to read and space to write.)

```
for( int pIndex = 0; pIndex < patches->size(); pIndex++ ){
    const Patch * patch = patches->get( pIndex );
    ostringstream file_name;
    // WARNING... not sure this is the correct naming convention...
    PLEASE VERIFY AND UPDATE THIS DOC.
    file_name << data_dir << "/p" << setw(5) << setfill('0') << pIndex;

    ofstream output( ceFileName, ios::out | ios::binary);
    if( !output ) {
        throw InternalError( string( "Component::saveMyData(): couldn't
            open file '" + file_name.str() + "'.",
            __FILE__, __LINE__ );
    }

    // Output data
    for( loop over data ) {
        output << data...
    }

    output.close();
} // end for pIndex

```



5 — Debugging

Debugging multi-processor software can be very difficult. Below are some hints on how to approach this.

5.1 Debugger

Use a debugger such as *GDB* to attach to the running program. Note, if you are an Emacs user, running GDB through Emacs' *gdb-mode* makes debugging even easier!

5.1.1 Serial Debugging

Whenever possible, debug *sus* running in serial (ie. no MPI). This is much easier for many reasons (though may also not be possible in many situations). To debug serially, use:

```
gdb sus <input_file>
```

5.1.2 Parallel Debugging

If it is necessary to use a parallel run of *sus* to debug, a few things must take place.

1) The helpful macro *WAIT_FOR_DEBUGGER()*; must be inserted into the code (*sus.cc*) just before the real execution of the components begins *ctl->run()*:

```
WAIT_FOR_DEBUGGER();  
ctl->run();
```

Recompile the code (will only take a few seconds), then run through MPI as normal.

Caveats

- Unless you know that the error occurs on a specific processor, you will need to attach a debugger to *every* process... so it behooves the developer to narrow the problem down to as few processors as possible.
- If you did know the processor, then the *WAIT_FOR_DEBUGGER()* will need to be placed inside an 'if' statement to check for the correct MPI Rank.
- Try to use only one node. If you need more 'processors', you (most likely) can double up on the same machine. If *sus* must span multiple nodes, then you will need to log into each node separately to attach debuggers (see below).

5.1.3 Running SUS

Run sus:

```
mpirun -np 8 -m mpihosts sus inputs/ARCHES/helium_1m.ups
```

Create a new terminal window for each processor you are using. (In the above case, 8 windows.)

You will notice that the output from sus will include lines (one for each processor used) like this:

```
updraft2:22793 waiting for debugger
```

Attach to each process in the following manner (where PID is the process id number (22793 in the above case)):

```
cd Uintah/bin/StandAlone
gdb -p <PID>
```

Within GDB, you will then need to break each process out of the WAIT loop in the following manner:

```
up 2
set wait=false
cont
```

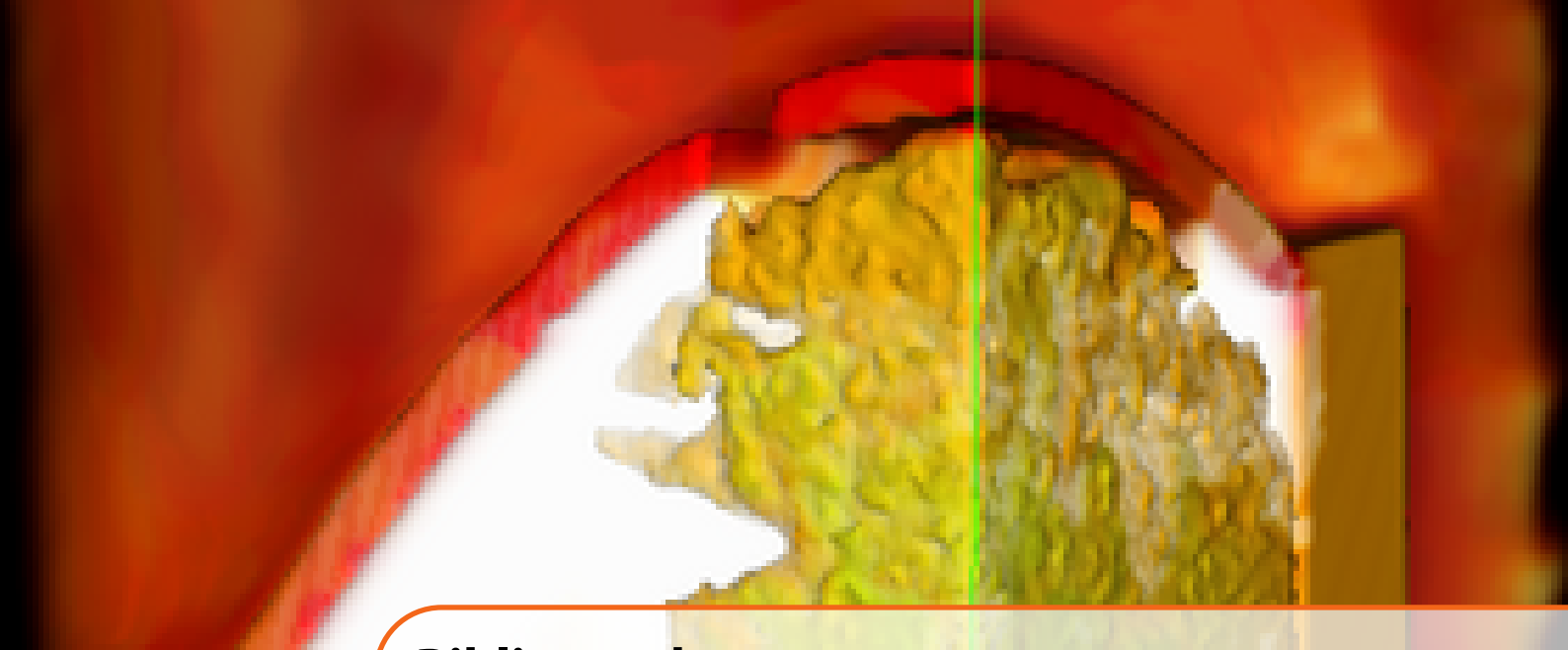
The above GDB commands must be run from each GDB session. Once all the WAITs are stopped, SUS will begin to run.

5.2 Memory Management

There are a number of tools that can help track down memory leaks in the code.

5.2.1 scinew

****TODO**** Write section.



Bibliography