# Domain Specific Embedded Languages in C++ Contributions to HPC



Joel Falcou

NumScale

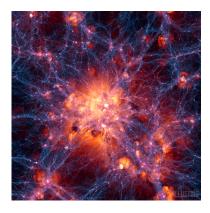
May 27, 2016



# The Paradigm Change in Science

### From Experiments to Simulations

- Simulations is now an integral part of the Scientific Method
- Scientific Computing enables larger, faster, more accurate Research
- Fast Simulation is Time Travel as scientific results are now more readily available



Local Galaxy Cluster Simulation - Illustris project

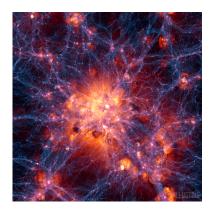
Computing is first and foremost a mainstream science tool



# The Paradigm Change in Science

#### The Parallel Hell

- Heat Wall: Growing cores instead of GHz
- Hierarchical and heterogeneous parallel systems are the norm
- The Free Lunch is over as hardware complexity rises faster than the average developer skills

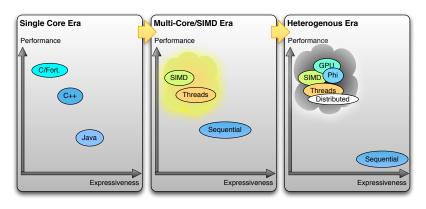


Local Galaxy Cluster Simulation - Illustris project

The real challenge in HPC is the Expressiveness/Efficiency War



# The Expressiveness/Efficiency War



As parallel systems complexity grows, the expressiveness gap turns into an ocean



# Designing tools for Scientific Computing

### **Objectives**

- I. Be non-disruptive
- 2. Domain driven optimizations
- 3. Provide intuitive API for the user
- 4. Support a wide architectural landscape
- 5. Be efficient



# Designing tools for Scientific Computing

### **Objectives**

- I. Be non-disruptive
- 2. Domain driven optimizations
- 3. Provide intuitive API for the user
- 4. Support a wide architectural landscape
- Be efficient

### Our Approach

- Design tools as C++ libraries (1)
- Design these libraries as Domain Specific Embedded Languages (DSEL) (2+3)
- Use Parallel Programming Abstractions as parallel components (4)
- Use Generative Programming to deliver performance (5)



# Talk Layout

Introduction

Abstractions & Efficiency

Experimental Results

Conclusion



# Why Parallel Programming Models?

#### Limits of regular tools

- Unstructured parallelism is error-prone
- Low level parallel tools are non-composable
- Contribute to the Expressiveness Gap



# Why Parallel Programming Models?

#### Limits of regular tools

- Unstructured parallelism is error-prone
- Low level parallel tools are non-composable
- Contribute to the Expressiveness Gap

#### Available Models

- Performance centric: P-RAM, LOG-P, BSP
- Pattern centric: Futures, Skeletons
- Data centric: HTA, PGAS



# Why Parallel Programming Models?

### Limits of regular tools

- Unstructured parallelism is error-prone
- Low level parallel tools are non-composable
- Contribute to the Expressiveness Gap

#### Available Models

- Performance centric: P-RAM, LOG-P, BSP
- Pattern centric: Futures, Skeletons
- Data centric: HTA, PGAS



# Parallel Skeletons [Cole 89]

#### **Principles**

- There are patterns in parallel applications
- Those patterns can be generalized in Skeletons
- Applications are assembled as a combination of such patterns

#### Functional point of view

- Skeletons are Higher-Order Functions
- Skeletons support a compositionnal semantic
- Applications become composition of state-less functions



### Parallel Skeletons [Cole 89]

#### **Principles**

- There are patterns in parallel applications
- Those patterns can be generalized in Skeletons
- Applications are assembled as a combination of such patterns

#### Classical Skeletons

- Data parallel: map, fold, scan
- Task parallel: par, pipe, farm
- More complex: Distribuable Homomorphism, Divide & Conquer, ...



# Domain Specific Embedded Languages

### Domain Specific Languages

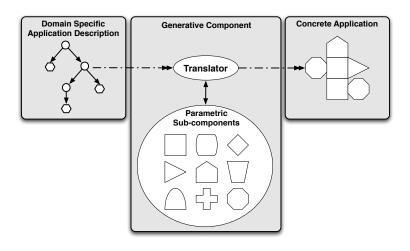
- Non-Turing complete declarative languages
- Solve a single type of problems
- Express what to do instead of how to do it
- E.g: SQL, Make, Matlab, ...

### From DSL to DSEL [Abrahams 2004]

- A DSL incorporates domain-specific notation, constructs, and abstractions as fundamental design considerations.
- A Domain Specific Embedded Languages (DSEL) is simply a library that meets the same criteria
- Generative Programming is one way to design such libraries



# Generative Programming [Eisenecker 97]





### Meta-programming as a Tool

#### Definition

Meta-programming is the writing of computer programs that analyse, transform and generate other programs (or themselves) as their data.

#### Meta-programmable Languages

- metaOCAML : runtime code generation via code quoting
- Template Haskell : compile-time code generation via templates
- C++: compile-time code generation via templates

### C++ meta-programming

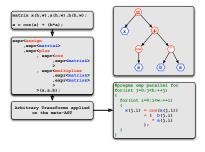
- Relies on the Turing-complete C++ TEMPLATE sub-language
- Handles types and integral constants at compile-time
- TEMPLATE classes and functions act as code quoting



### The Expression Templates Idiom

### **Principles**

- Relies on extensive operator overloading
- Carries semantic information around code fragment
- Introduces DSLs without disrupting dev. chain



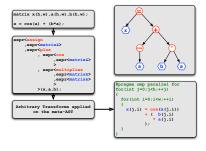
General Principles of Expression Templates



### The Expression Templates Idiom

### **Advantages**

- Generic implementation becomes self-aware of optimizations
- API abstraction level is arbitrary high
- Accessible through high-level tools like Boost.Proto



General Principles of Expression Templates



#### Our Contributions

### Our Strategy

- Applies DSEL generation techniques to parallel programming
- Maintains low cost of abstractions through meta-programming
- Maintains abstraction level via modern library design

#### Our contributions

Tools	Pub.	Scope	Applications
Quaff	ParCo'06	MPI Skeletons	Real-time 3D reconstruction
SkellBE	PACT'08	Skeleton on Cell BE	Real-time Image processing
BSP++	IJPP'12	MPI/OpenMP BSP	Bioinformatics, Model Checking
$NT^2$	JPDC'14	Data Parallel Matlab	Fluid Dynamics, Vision



#### Second Look at our Contributions

#### **Development Limitations**

- DSELs are mostly tied to the domain model
- Architecture support is often an afterthought
- Extensibility is difficult as many refactoring are required per architecture
- Example: No proper way to support GPUs with those implementation techniques



#### Second Look at our Contributions

#### **Development Limitations**

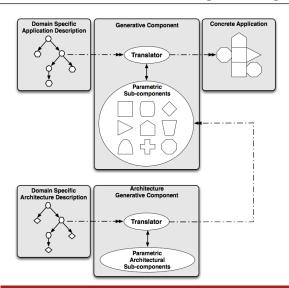
- DSELs are mostly tied to the domain model
- Architecture support is often an afterthought
- Extensibility is difficult as many refactoring are required per architecture
- Example: No proper way to support GPUs with those implementation techniques

#### Proposed Method

- Extends Generative Programming to take this architecture into account
- Provides an architecture description DSEL
- Integrates this description in the code generation process



# Architecture Aware Generative Programming





# Software refactoring

Tools	Issues	Changes
Quaff	Raw skeletons API	Re-engineered as part of NT <sup>2</sup>
<del>SkellBE</del>	Too architecture specific	Re-engineered as part of $NT^2$
BSP++	Integration issues	Integrate hybrid code generation
$NT^2$	Not easily extendable	Integrate Quaff Skeleton models
Boost.SIMD	-	Side product of $NT^2$ restructuration

#### Conclusion

- Skeletons are fine as parallel middleware
- Model based abstractions are not high level enough
- For low level architectures, the simplest model is often the best

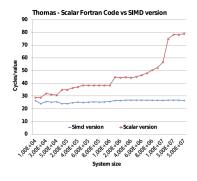


#### Boost.SIMD

#### Pierre Estérie PHD 2010-2014

#### **Principles**

- Provides simple C++ API over SIMD extensions
- Supports every Intel and PPC instructions sets
- Fully integrates with modern C++ idioms



Sparse Tridiagonal Solver - collaboration with M. Baboulin and Y. wang



# Talk Layout

Introduction

**Abstractions & Efficiency** 

**Experimental Results** 

Conclusion



Pierre Estérie PHD 2010-2014

### NT<sup>2</sup> as a Scientific Computing Library

- Provides a simple, Matlab-like interface for users
- Provides high-performance computing entities and primitives
- Is easily extendable

#### Components

- Uses Boost.SIMD for in-core optimizations
- Uses recursive parallel skeletons
- Supports task parallelism through Futures



#### **Principles**

- table<T, S> is a simple, multidimensional array object that exactly mimics MATLAB array behavior and functionalities
- 500+ functions usable directly either on table or on any scalar values as in MATLAB



### **Principles**

- table<T, S> is a simple, multidimensional array object that exactly mimics MATLAB array behavior and functionalities
- 500+ functions usable directly either on table or on any scalar values as in MATLAB

#### How does it works

Take a .m file, copy to a .cpp file



### **Principles**

- table<T, S> is a simple, multidimensional array object that exactly mimics MATLAB array behavior and functionalities
- 500+ functions usable directly either on table or on any scalar values as in MATLAB

#### How does it works

- Take a .m file, copy to a .cpp file
- Add #include <nt2/nt2.hpp> and do cosmetic changes



### **Principles**

- table<T, S> is a simple, multidimensional array object that exactly mimics MATLAB array behavior and functionalities
- 500+ functions usable directly either on table or on any scalar values as in MATLAB

#### How does it works

- Take a .m file, copy to a .cpp file
- Add #include <nt2/nt2.hpp> and do cosmetic changes
- Compile the file and link with libnt2.a

#### NT2 - From MATLAB to C++

#### MATLAB code

```
A1 = 1:1000;
A2 = A1 + randn(size(A1));
X = lu(A1*A1');

rms = sqrt( sum(sqr(A1(:) - A2(:))) / numel(A1) );

NT<sup>2</sup> code

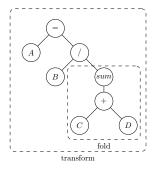
table < double > A1 = _(1.,1000.);
table < double > A2 = A1 + randn(size(A1));
table < double > X = lu( mtimes(A1, trans(A1) );

double rms = sqrt( sum(sqr(A1(_) - A2(_))) / numel(A1) );
```



### Parallel Skeletons extraction process

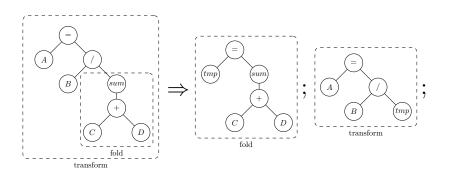
$$A = B / sum(C+D);$$





# Parallel Skeletons extraction process

$$A = B / sum(C+D);$$





### From data to task parallelism

Antoine Tran Tan PHD, 2012-2015

### Limits of the fork-join model

- Synchronization cost due to implicit barriers
- Under-exploitation of potential parallelism
- Poor data locality and no inter-statement optimization



### From data to task parallelism

Antoine Tran Tan PHD, 2012-2015

#### Limits of the fork-join model

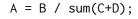
- Synchronization cost due to implicit barriers
- Under-exploitation of potential parallelism
- Poor data locality and no inter-statement optimization

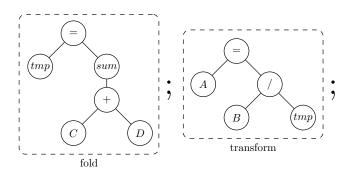
#### Skeletons from the Future

- Adapt current skeletons for taskification
- Use Futures (STD or HPX) to automatically pipeline
- Derive a dependency graph between statements



### Parallel Skeletons extraction process - Take 2

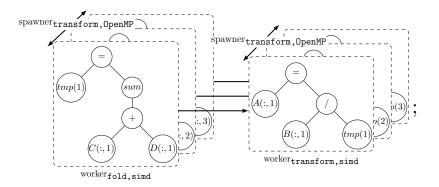






### Parallel Skeletons extraction process - Take 2

$$A = B / sum(C+D);$$





## Motion Detection

#### Lacassagne et al., ICIP 2009

- Sigma-Delta algorithm based on background substraction
- Use local gaussian model of lightness variation to detect motion
- Challenge: Very low arithmetic density
- Challenge: Integer-based implementation with small range





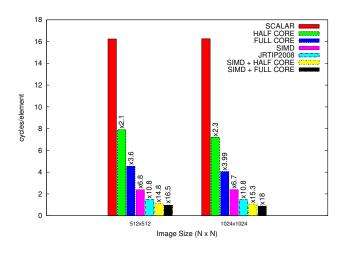


#### Motion Detection

```
table < char > sigma_delta( table < char > & background
                        . table<char> const& frame
                          table<char>& variance
  // Estimate Raw Movement
  background = selinc( background < frame
                  , seldec(background > frame, background)
                 );
  table < char > diff = dist(background, frame):
  // Compute Local Variance
  table < char > sig3 = muls(diff.3):
  var = if else( diff != 0
                . selinc( variance < sig3
                       . seldec( var > sig3. variance)
                , variance
  // Generate Movement Label
  return if_zero_else_one( diff < variance );</pre>
```



## Motion Detection





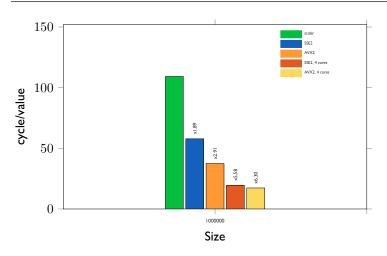
## NT<sup>2</sup> Code



# NT<sup>2</sup> Code with loop fusion

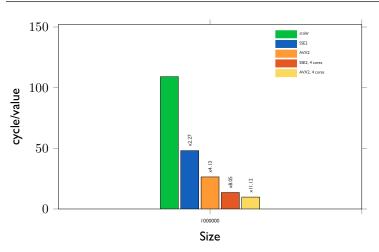


#### Performance





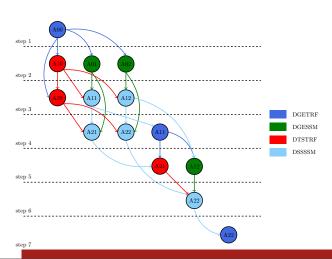
Performance with loop fusion/futurisation





# LU Decomposition

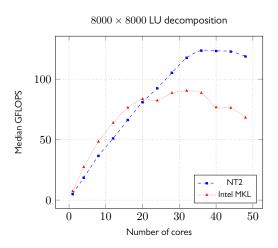
#### Algorithm





# LU Decomposition

#### Performance





# Talk Layout

Introduction

**Abstractions & Efficiency** 

**Experimental Results** 

Conclusion



#### Conclusion

# Parallel Computing for Scientist

- Software Libraries built as Generic and Generative components can solve a large chunk of parallelism related problems while being easy to use.
- Like regular language, DSEL needs informations about the hardware system
- Integrating hardware descriptions as Generic components increases tools portability and re-targetability



## Conclusion

## Parallel Computing for Scientist

- Software Libraries built as Generic and Generative components can solve a large chunk of parallelism related problems while being easy to use.
- Like regular language, DSEL needs informations about the hardware system
- Integrating hardware descriptions as Generic components increases tools portability and re-targetability

#### Our Achievements

- A new method for parallel software development
- Efficient libraries working on large subset of hardware
- High level of performances across a wide application spectrum



# Perspectives

#### DSEL as C++ first class idiom

- Build partial evaluation into the language
- Ease transition between regular and meta C++
- Mid-term Prospect: metaOCAML like quoting for C++

## DSEL and compilers relationship

- C++ DSEL hits a limit on their applicability
- Compilers often lack high level informations for proper optimization
- Mid-term Prospect: Hybrid library/compiler approaches for DSEL

# Thanks for your attention