# A Discussion of Selected Vienna-Libraries for Computational Science

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C++Now, May 15th, 2013

Part 0: What is this all about?

## Libraries

ViennaCL

ViennaData

ViennaFEM

ViennaGrid

ViennalPD

ViennaMath

ViennaMesh

# **Applications**

ViennaMOS

ViennaProfiler

ViennaSHE

ViennaWD

ViennaX

#### Covered In This Talk

Requirements in Computational Science

**GPU Computing** 

Providing High-Level Interfaces

Avoiding Monolithic Code

Abstractions: From Math to Code (and back)

#### **Not Covered**

C++11

Maximize use of Boost

# The Computational Scientist

## A Strange Animal

Goal is science, not to execute software

Codes seldomly designed for 'large scale' upfront

Scientists receive software training from scientists

Performance vs. portability and maintainability

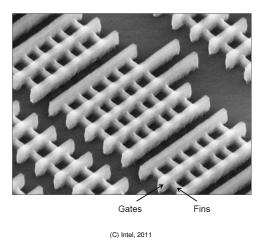
Run code on clusters (headless, old software stack, ...)

Come and go

Basili *et al.*, Understanding the High-Performance-Computing Community: A Software Engineer's Perspective. IEEE Software, 2008.

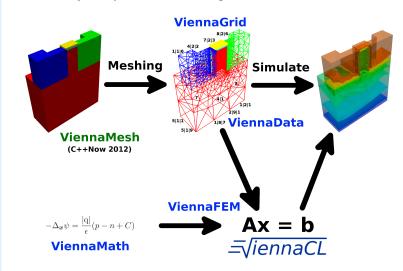
## **Simulation Flow**

# The Many Steps in Simulating a FinFET



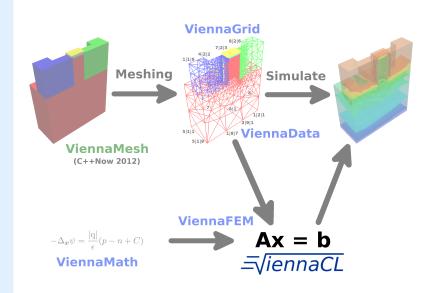
## Simulation Flow

## The Many Steps in Simulating a FinFET



Part 1: ViennaCL

## **Simulation Flow**



## Consider Existing CPU Code (Boost.uBLAS)

```
using namespace boost::numeric::ublas;
matrix<double> A(1000, 1000);
vector<double> x(1000), y(1000);
/* Fill A, x, y here */
double val = inner_prod(x, y);
y += 2.0 * x;
A += val * outer_prod(x, y);
x = solve(A, y, upper_tag()); // Upper tri. solver
std::cout << " 2-norm: " << norm_2(x) << std::endl;
std::cout << "sup-norm: " << norm_inf(x) << std::endl</pre>
```

#### Previous Code Snippet Rewritten with ViennaCL

```
using namespace viennacl;
using namespace viennacl::linalg;
matrix<double> A(1000, 1000);
vector<double> x(1000), y(1000);
/* Fill A, x, y here */
double val = inner_prod(x, y);
y += 2.0 * x;
A += val * outer_prod(x, y);
x = solve(A, y, upper_tag()); // Upper tri. solver
std::cout << " 2-norm: " << norm_2(x) << std::endl;
std::cout << "sup-norm: " << norm_inf(x) << std::endl</pre>
```

#### ViennaCL in Addition Provides Iterative Solvers

```
using namespace viennacl;
using namespace viennacl::linalg;
compressed matrix<double> A(1000, 1000);
vector<double> x(1000), y(1000);
/* Fill A, x, y here */
x = solve(A, y, cq_tag()); // Conjugate
   Gradients
x = solve(A, y, bicgstab_tag()); // BiCGStab solver
x = solve(A, y, qmres_tag()); // GMRES solver
```

No Iterative Solvers Available in Boost.uBLAS...

## Thanks to Interface Compatibility

```
using namespace boost::numeric::ublas;
using namespace viennacl::linalg;
compressed_matrix<double> A(1000, 1000);
vector<double> x(1000), y(1000);
/* Fill A, x, y here */
x = solve(A, y, cq_tag()); // Conjugate
   Gradients
x = solve(A, y, bicgstab_tag()); // BiCGStab solver
x = solve(A, y, gmres_tag()); // GMRES solver
```

## Code Reuse Beyond GPU Borders

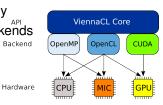
```
Eigen http://eigen.tuxfamily.org/
MTL 4 http://www.mtl4.org/
```

## Generic CG Implementation (Sketch)

```
for (unsigned int i = 0; i < tag.max_iterations(); ++</pre>
   i)
  tmp = viennacl::linalq::prod(A, p);
  alpha = ip_rr / inner_prod(tmp, p);
  result += alpha * p;
  residual -= alpha * tmp;
  new_ip_rr = inner_prod(residual, residual);
  if (new_ip_rr / norm_rhs_squared < tag.tolerance())</pre>
    break:
  beta = new_ip_rr / ip_rr;
  ip_rr = new_ip_rr;
  p = residual + beta * p;
```

#### **About**

High-level linear algebra C++ library
OpenMP, OpenCL, and CUDA backends
Header-only
Multi-platform



#### Dissemination

Free Open-Source MIT (X11) License http://viennacl.sourceforge.net/ 50-100 downloads per week

## **Design Rules**

Reasonable default values Compatible to Boost.uBLAS whenever possible In doubt: clean design over performance

# **Basic Types**

scalar, vector
matrix, compressed\_matrix, coordinate\_matrix, ell\_matrix,
hyb\_matrix

#### **Data Initialization**

```
std::vector<double> std_x(100);
ublas::vector<double> ublas_x(100);
viennacl::vector<double> vcl_x(100);

for (size_t i=0; i<100; ++i)
   // std_x[i] = rand(); // (1)
   // ublas_x[i] = rand(); // (2)
   vcl_x[i] = rand(); // (3)</pre>
```

(3) is slowest by orders of magnitude!

## **Basic Types**

scalar, vector
matrix, compressed\_matrix, coordinate\_matrix, ell\_matrix,
hyb\_matrix

#### **Data Initialization**

Using viennacl::copy()

## **Basic Types**

scalar, vector
matrix, compressed\_matrix, coordinate\_matrix, ell\_matrix,
hyb\_matrix

#### **Data Initialization**

Using viennacl::copy()

```
std::vector<std::vector<double> > std_A;
ublas::matrix<double> ublas_A;
viennacl::matrix<double> vcl_A;

/* setup of std_A and ublas_A omitted */
viennacl::copy(std_A, vcl_A); // CPU to GPU
viennacl::copy(vcl_A, ublas_A); // GPU to CPU
```

Iterator concept doesn't quite work on accelerators

#### **Vector Addition**

```
x = y + z;
```

Temporaries are costly (particularly on GPUs)

## **Expression Templates**

Limited expansion

Map to a set of predefined kernels

```
vector_expression<vector<T>, op_plus, vector<T> >
operator+(vector<T> & v, vector<T> & w) { ... }

vector::operator=(vector_expression<...> const & e)
    {
    viennacl::linalg::avbv(*this, 1,e.lhs(), 1,e.rhs()
    );
}
```

#### **Vector Addition**

```
// x = v + z
void avbv(...) {
  switch (active handle id(x)) {
    case MAIN_MEMORY:
      host based::avbv(...);
      break;
    case OPENCL_MEMORY:
      opencl::avbv(...);
      break;
    case CUDA_MEMORY:
      cuda::avbv(...);
      break;
    default:
      raise error();
```

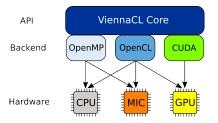
Memory buffers can switch memory domain at runtime

## Memory Buffer Migration

```
vector<double> x = zero_vector<double>(42);

memory_types src_memory_loc = memory_domain(x);
switch_memory_domain(x, MAIN_MEMORY);

/* work on x in main memory here */
switch_memory_domain(x, src_memory_loc);
```



# Generalizing compute kernels

```
// x = v + z
__kernel void avbv(
  double * x,
  double * V,
  double * z, uint size)
size_t i = get_global_id(0);
for (; i<size; i += get_global_size())</pre>
  x[i] = y[i] + z[i];
```

## Generalizing compute kernels

```
// x = a * v + b * z
__kernel void avbv(
  double * x,
  double a,
  double * y,
  double b,
  double * z, uint size)
size_t i = get_global_id(0);
for (; i<size; i += get_global_size())</pre>
  x[i] = a * y[i] + b * z[i];
```

## Generalizing compute kernels

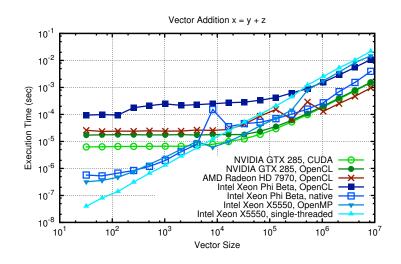
```
// x[4:8] = a * y[2:6] + b * z[3:7]
__kernel void avbv(
  double * x, uint off_x,
  double a,
  double * y, uint off_y,
  double b,
  double * z, uint off_z, uint size)
size_t i = get_global_id(0);
for (; i<size; i += get_global_size())</pre>
  x[off_x + i] = a * y[off_y + i] + b * z[off_z + i]
     1;
```

## Generalizing compute kernels

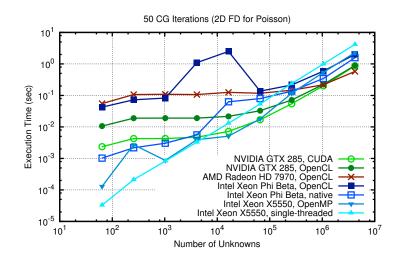
```
// x[4:2:8] = a * y[2:2:6] + b * z[3:2:7]
  kernel void avbv(
    double * x, uint off x, uint inc x,
    double a.
    double * y, uint off_y, uint inc_y,
    double b.
    double * z, uint off_z, uint inc_z, uint size)
  size_t i = get_global_id(0);
  for (; i<size; i += get_global_size())</pre>
    x[off_x + i * inc_x] = a * y[off_y + i * inc_y]
                          + b * z[off z + i * inc z];
}
```

No penalty on NVIDIA GPUs because FLOPs are for free

## **Benchmarks**



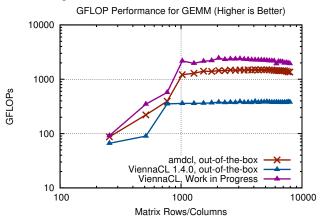
## **Benchmarks**



## **Benchmarks**

## Matrix-Matrix Multiplication

#### Autotuning environment

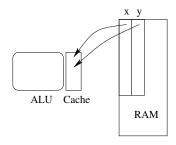


(AMD Radeon HD 7970, single precision)

## Expression Templates are Not Enough

#### Consider

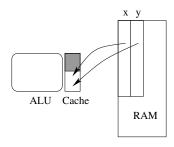
#### Suboptimal performance with almost any library



## Expression Templates are Not Enough

#### Consider

#### Suboptimal performance with almost any library



## Expression Templates are Not Enough

#### Consider

```
u = x + y;
v = x - y;
```

Suboptimal performance with almost any library

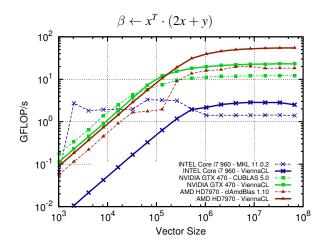
## **OpenCL Kernel Generation**

Separate temporary avoidance from operation execution

```
custom_operation op;
op.add( u = x + y );
op.add( v = x - y );
op.execute();  // OpenCL JIT
```

Fully transparent kernel fusion scheduled for release 1.5.0

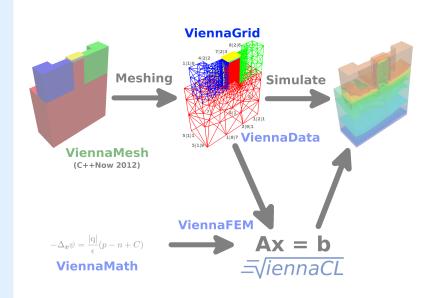
#### **Benchmark Results**



Tillet et al., HotPar '13

Part 2: ViennaGrid

## **Simulation Flow**



# What about Boost.Geometry?

## A Quick Look at Boost.Geometry

Concepts: Point, Segment, Linestring, Box, etc.

Algorithms: distance(), intersects(), convex\_hull(),
etc.

Central entity: Point

```
int a[3] = {1, 2, 3};
int b[3] = {2, 3, 4};

double d = boost::geometry::distance(a, b);
```

## Why Boost.Geometry is Not Enough

What about 3D objects? Tetrahedra? Hexahedra?

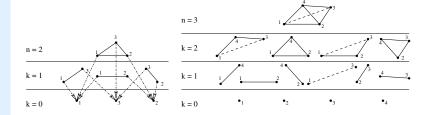
Traversal of boundary objects missing

No storage facilities for many objects

# *n*-cell Concept

## Concept of *n-cell*

Sub-k-cells of an n-cell



## Separation of Geometry and Topology

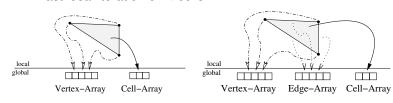
Geometry: Euclidian space, coordinate system

Topology: Connection between points (lines, triangles, etc.)

## **ViennaGrid Datastructure**

## **Datastructure Requirements**

Don't store boundary *k*-cells unnecessarily Fast local iteration on *k*-cells



		Amount	Mem/Obj.	Total Mem.
	Vertices	4913	24 B	115 KB
	Edges	31024	16 B	485 KB
	Facets	50688	48 B	2376 KB
	Cells	24576	112 B	2688 KB
	Total			5664 KB
		Amount	Mem/Obj.	Total Mem.
	Vertices	4913	24 B	115 KB
		_		2.175

## *n*-cell Implementation

## Implementation of element\_t

Recursive inheritance from boundary layer of dimension n-1 Tag dispatching to enable/disable topological layers

```
element_t
boundary_ncell_layer<n-1> ID
boundary_ncell_layer<0>
boundary_ncell_layer<0>
```

```
template <typename ConfigType, typename ElementTag>
class element_t :
   public boundary_ncell_layer<ConfigType, ElementTag, ElementTag::
        dim-1>,
   public result_of::element_id_handler<ConfigType, ElementTag>::
        type
```

## **Domain Configuration**

## Top Level Configuration of Domain

Mostly a collection of tags

Predefined configuration classes for common cases ViennaGrid 1.0.x: Supplemented by global customizations

Work in progress: Everything in a single config class

#### ViennaGrid User API

## User API Design Goals

STL-style, reuse conventions

Allow index-based traversal

Avoid common C++ pitfalls (e.g. template member functions)

Ranges provide iterators over *n*-cells Extendible

## ViennaGrid User API

## User API Design Goals

STL-style, reuse conventions

Allow index-based traversal

Avoid common C++ pitfalls (e.g. template member functions)

```
//iteration over all vertices in the domain:

for (std::size_t i = 0; i < ncells<0>(domain).size(); ++i)
{
   // do something with ncells<0>(domain)[i] here
}
```

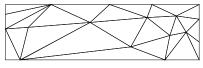
#### OpenMP friendly

.size() sometimes known at compiletime!

#### ViennaGrid User API

## User API Design Goals

Boundary iteration: k < nCoboundary iteration: k > n



Coboundary information not a-priori available in datastructure Built and cached at first request

## **ViennaGrid Features**

#### Other Features

Segments

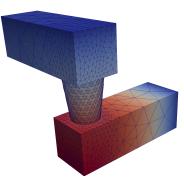
I/O: VTK, various mesh-format

Voronoi information

Refinement

## Work in Progress

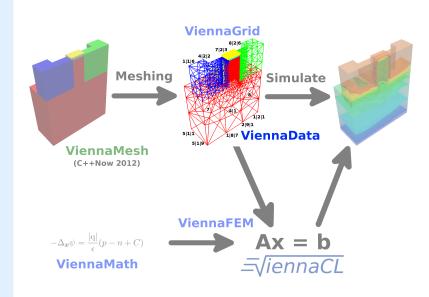
PLC, hybrid meshes, multigrid



## **Contents**

Part 3: ViennaData

## **Simulation Flow**



## **Data Storage**

## Plain Object-Oriented Approach?

```
struct Triangle {
  PointType a, b, c;
  bool is_on_boudary;
  double rho; }; //e.g. specific mass
```

#### **Pros and Cons**

Data is directly stored with the object Each (!) triangle carries a boolean flag Reusability reduced ⇒ better rename to TriangleWithMaterial

# Monolithic, don't do this!

## **Data Storage**

#### Store Data with Mesh:

```
class Mesh {
  vector<Triangle> triangles;
  std::map<size_t, bool> boudary_map; //no memory
     wasted
  vector<double> rho_for_triangles; };
```

#### **Pros and Cons**

Triangle is still reusable

Mesh class has to handle data storage complexity

Each additional data requires a change of Mesh class

Mesh object has to be passed to all modules

## Monolithic, don't do this!

## Approach by ViennaData

Introduction of a hidden data container

Data is stored in a map-like manner using keys

```
//generic interface:
viennadata::access<KeyType, ValueType>(key)(obj);

//boundary flag and specific mass retrieval for triangle:
bool on_boundary = access<BoundaryKeyType, bool>(boundary_key)(
    triangle);
double    rho = access<MassKeyType, double>( mass_key)(
    triangle);
```

#### **Pros and Cons**

- + Triangle and Mesh are still reusable
- + Can be used with arbitrary objects (third-party libraries)
- + Unified interface for data access
- Global state

#### Generic Interface

```
viennadata::access<KeyType, ValueType>(key)(obj);
```

## Default Storage Scheme

If nothing is known about the object:

```
std::map<ObjectType *,
    std::map<KeyType, ValueType> >
```

The compiler creates such a map for each of the following:

```
access<long, double>(42) (triangle);
access<char, double>('b') (triangle);
access<std::string, double>("mass") (triangle);
access<std::string, double>("mass") (vertex);
```

#### Performance Considerations

 $\mathcal{O}(\log N) + \mathcal{O}(\log K)$  access time

N ... objects of same type (e.g. triangles)

 $K \dots$  keys of same type

Usually too slow in a high-performance setting

#### Can We Do Better?

In general: No

In certain situations: Yes

## Type-based Key Dispatch

Only one key per type  $\mathcal{O}(\log N)$  access time

```
access<mass_key, double>(mass_key())(triangle);

// or:
access<mass_key, double>()(triangle);
```

#### Internal Datastructure

```
std::map<ObjectType *, ValueType>
```

Transparent to user

One line of code for activation

## Numeric IDs for Objects

- Only one key per type
- $\mathcal{O}(K)$  access time
- $\mathcal{O}(1)$  access time with type-based key dispatch

```
access<std::string, double>("mass")(triangle);

// with type-based key dispatch:
access<mass_key, double>()(triangle);
```

#### Internal Datastructure

```
std::vector<ValueType>
```

Transparent to user

One line of code for activation

Overload generic id() accessor

## Benchmarking ViennaData

ID-based access to data via ViennaData (class LightWeight) OOP-style storage in classes with payload

```
template <size_t N>
struct FatClass {
  double data;
  char payload[N];
};
```

	10 <sup>3</sup> Objects (us)	10 <sup>6</sup> Objects (ms)
LightWeight	4	5
FatClass<10>	1.3	4
FatClass<100>	2.1	11
FatClass<1000>	2.5	11

#### Other Routines

Not further addressed:

```
viennadata::copy<KeyType, ValueType>(key)(
    src_obj, dst_obj);
viennadata::move<KeyType, ValueType>(key)(
    src_obj, dst_obj);
viennadata::find<KeyType, ValueType>(key)(obj);
viennadata::erase<KeyType, ValueType>(key)(obj);
viennadata::erase<KeyType, all>(key)(obj);
viennadata::erase<all, ValueType>(key)(obj);
```

#### **Pitfalls**

Inheritance

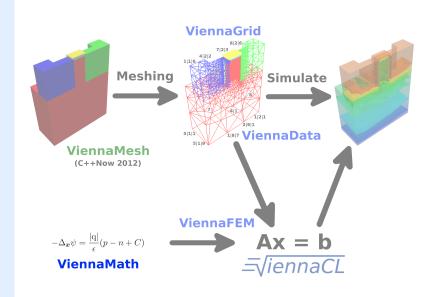
Limited lifetime of objects

Compilation units with different configuration

## **Contents**

Part 4: ViennaMath

## **Simulation Flow**



## A Symbolic Math Library in C++

Symbolic evaluation and manipulation of math expressions
Unified run time and compile time interface
Differentiation and integration provided
LATEX output

## **Example Usage**

```
variable x(0);
variable y(1);
variable z(2);
expr f = x + y - z;
expr g = f * f;
eval(f, make_vector(1.0, 2.0, 4.0)); // returns
-1.0
eval(g, make_vector(1.0, 2.0, 4.0)); // returns
1.0
```

## A Symbolic Math Library in C++

Symbolic evaluation and manipulation of math expressions
Unified run time and compile time interface
Differentiation and integration provided
LATEX output

## **Example Usage**

```
ct_variable<0> x;
ct_variable<1> y;
ct_variable<2> z;
ct_constant<1> c1;
ct_constant<2> c2;
ct_constant<4> c4;
eval(x + y - z, make_vector(c1, c2, c4)); //
returns -1
```

#### Compile Time Evaluation

## Substitution and Differentiation (Run Time)

```
variable x(0);
variable y(1);
variable z(2);
substitute(x, y, x*y + z); // returns y*y+z
diff(x*y + z, x); // returns y
```

## Substitution and Differentiation (Compile Time)

```
ct_variable<0> x;
ct_variable<1> y;
ct_variable<2> z;
substitute(x, y, x*y + z); // returns y*y+z
diff(x*y + z, x); // returns y
```

## Numerical Integration (Run Time): $\int_0^1 x^2 dx$

# Analytical Integration (Compile Time): $\int_0^1 x^2 dx$ , $\int_0^1 \int_0^{1-x} xy dxdy$

## **Function Symbols**

Represent a function (not evaluable)

## **Differential Operators**

Gradient, Divergence

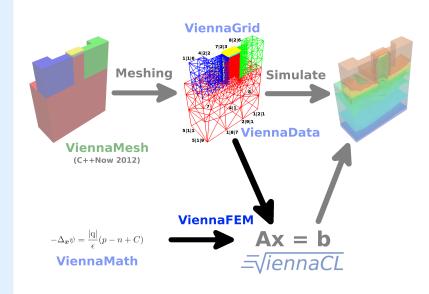
## Symbolic Integration Domain

Specify actual integration domain and integration variables *later* 

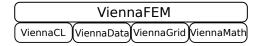
## **Contents**

Part 5: ViennaFEM

## **Simulation Flow**



## Simulation Flow



## Library-Centric Design

ViennaCL for linear solver
ViennaData for data storage
ViennaGrid for mesh handling
ViennaMath for symbolic math

#### Addresses Come&Go in Academia

Focus on one package

No in-depth knowledge of all packages required

Additional emphasis on good interfaces

## **ViennaFEM**

#### ViennaGrid deals with the Mesh

```
DomainType my_domain;
viennagrid::io::netgen_reader my_reader;
my_reader(my_domain, "mystructure.mesh");
```

#### Equation specification via ViennaMath: $\Delta u = -1$

```
equation poisson_eq = viennamath::make_equation(laplace(u), -1);
```

## Assembly via ViennaFEM

## Linear solver provided by ViennaCL

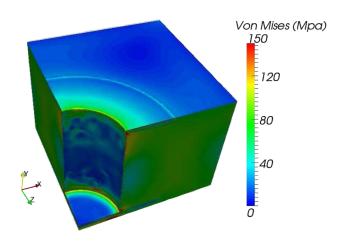
## **ViennaFEM**

## Lame equation for Linear Elasticity

$$\int_{\Omega} \varepsilon(u) : \sigma(v) \, d\Omega = \int_{\Omega} F \cdot v \, d\Omega \quad \forall v \in \mathcal{V}$$

With  $F = (0, 0, 1)^{T}$ :

## Solving Lame's Equation for a TSV



## Summary

## Software Packages

ViennaCL: GPU-accelerated Linear Algebra

ViennaData: Generic Data Storage

ViennaFEM: Modular High-Level Finite Element Package

ViennaGrid: Generic Mesh Datastructure

ViennaMath: Symbolic Math Kernel

http://vienna{cl,data,fem,grid,math}@sourceforge.ne

## Design Philosophy

Orthogonal Software Design Convenient High-Level User API Avoid Unneeded Dependencies Free Open Source (MIT License)