

VexCL

Vector Expression Template Library for OpenCL

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VexCL: Vector expression template library for OpenCL

- Created for ease of C++ based OpenCL development.
- The source code is publicly available¹ under MIT license.
- *This is not a C++ bindings library!*

1 Motivation

2 Basic interface

3 Kernel builder

4 Performance

5 Implementation details

6 Conclusion

¹<https://github.com/ddemidov/vexcl>

Hello OpenCL: vector sum

Vector sum

- A , B , and C are large vectors.
- Compute $C = A + B$.

Overview of OpenCL solution

- ➊ Initialize OpenCL context on supported device.
- ➋ Allocate memory on the device.
- ➌ Transfer input data to device.
- ➍ Run your computations on the device.
- ➎ Get the results from the device.

Hello OpenCL: vector sum

1. Query platforms

```
1 std::vector<cl::Platform> platform;
2 cl::Platform::get(&platform);
3
4 if ( platform.empty() ) {
5     std::cerr << "OpenCL platforms not found." << std::endl;
6     return 1;
7 }
```

Hello OpenCL: vector sum

2. Get first available GPU device

```

8 cl :: Context context;
9 std :: vector<cl::Device> device;
10 for(auto p = platform.begin(); device.empty() && p != platform.end(); p++) {
11     std :: vector<cl::Device> pldev;
12     try {
13         p->getDevices(CL_DEVICE_TYPE_GPU, &pldev);
14         for(auto d = pldev.begin(); device.empty() && d != pldev.end(); d++) {
15             if (!d->getInfo<CL_DEVICE_AVAILABLE>()) continue;
16             device.push_back(*d);
17             context = cl::Context(device);
18         }
19     } catch(...) {
20         device. clear ();
21     }
22 }
23 if (device. empty ()) {
24     std :: cerr << "GPUs not found." << std::endl;
25     return 1;
26 }
```

Hello OpenCL: vector sum

3. Create kernel source

```

27 const char source[] =
28     "kernel void add(\n"
29     "    uint n,\n"
30     "    global const float *a,\n"
31     "    global const float *b,\n"
32     "    global float *c\n"
33     ")\n"
34     "{\n"
35     "    uint i = get_global_id(0);\n"
36     "    if (i < n) {\n"
37     "        c[i] = a[i] + b[i];\n"
38     "    }\n"
39     "}";

```

Hello OpenCL: vector sum

4. Compile kernel

```
40 cl::Program program(context, cl::Program::Sources(  
41     1, std::make_pair(source, strlen(source))  
42 ));  
43 try {  
44     program.build(device);  
45 } catch (const cl::Error&) {  
46     std::cerr  
47         << "OpenCL compilation error" << std::endl  
48         << program.getBuildInfo<CL_PROGRAM_BUILD_LOG>(device[0])  
49         << std::endl;  
50     return 1;  
51 }  
52 cl::Kernel add_kernel = cl::Kernel(program, "add");
```

5. Create command queue

```
53 cl::CommandQueue queue(context, device[0]);
```

Hello OpenCL: vector sum

6. Prepare input data, transfer it to device

```
54 const unsigned int N = 1 << 20;  
55 std :: vector<float> a(N, 1), b(N, 2), c(N);  
56  
57 cl :: Buffer A(context, CL_MEM_READ_ONLY | CL_MEM_COPY_HOST_PTR,  
58     a.size () * sizeof(float ), a.data());  
59  
60 cl :: Buffer B(context, CL_MEM_READ_ONLY | CL_MEM_COPY_HOST_PTR,  
61     b.size () * sizeof(float ), b.data());  
62  
63 cl :: Buffer C(context, CL_MEM_READ_WRITE,  
64     c.size () * sizeof(float ));
```

Hello OpenCL: vector sum

7. Set kernel arguments

```
65 add_kernel.setArg(0, N);
66 add_kernel.setArg(1, A);
67 add_kernel.setArg(2, B);
68 add_kernel.setArg(3, C);
```

8. Launch kernel

```
69 queue.enqueueNDRangeKernel(add_kernel, cl::NullRange, N, cl::NullRange);
```

9. Get result back to host

```
70 queue.enqueueReadBuffer(C, CL_TRUE, 0, c.size() * sizeof(float), c.data());
71 std::cout << c[42] << std::endl; // Should get '3' here.
```

Hello VexCL: vector sum

This is much shorter!

```
1 std :: cout << 3 << std::endl;
```

Hello VexCL: vector sum

Get all available GPUs

```
1 vex::Context ctx( vex::Filter ::Type(CL_DEVICE_TYPE_GPU) );
2 if ( !ctx.size() ) {
3     std::cerr << "GPUs not found." << std::endl;
4     return 1;
5 }
```

Prepare input data, transfer it to device

```
6 std::vector<float> a(N, 1), b(N, 2), c(N);
7 vex::vector<float> A(ctx.queue(), a);
8 vex::vector<float> B(ctx.queue(), b);
9 vex::vector<float> C(ctx.queue(), N);
```

Launch kernel, get result back to host

```
10 C = A + B;
11 vex::copy(C, c);
12 std::cout << c[42] << std::endl;
```

1 Motivation

2 Basic interface

- Device selection
- Vector arithmetic
- Reductions
- User-defined functions
- Sparse matrix – vector products
- Stencil convolutions
- Multivectors & multiexpressions

3 Kernel builder

4 Performance

5 Implementation details

6 Conclusion

Device selection

- Multi-device and multi-platform computations are supported.
- VexCL context is initialized from combination of device filters.
- Device filter is a boolean functor acting on `const cl::Device&`.

Initialize VexCL context on selected devices

```
1 vex::Context ctx( vex::Filter :: All );
```



Device selection

- Multi-device and multi-platform computations are supported.
- VexCL context is initialized from combination of device filters.
- Device filter is a boolean functor acting on `const cl::Device&`.

Initialize VexCL context on selected devices

```
1 vex::Context ctx( vex::Filter ::Type(CL_DEVICE_TYPE_GPU) );
```



Device selection

- Multi-device and multi-platform computations are supported.
- VexCL context is initialized from combination of device filters.
- Device filter is a boolean functor acting on `const cl::Device&`.

Initialize VexCL context on selected devices

```
1 vex::Context ctx(  
2     vex::Filter ::Type(CL_DEVICE_TYPE_GPU) &&  
3     vex::Filter ::Platform("AMD")  
4 );
```



Device selection

- Multi-device and multi-platform computations are supported.
- VexCL context is initialized from combination of device filters.
- Device filter is a boolean functor acting on `const cl::Device&`.

Initialize VexCL context on selected devices

```
1 vex::Context ctx(  
2     vex::Filter ::Type(CL_DEVICE_TYPE_GPU) &&  
3     []( const cl::Device &d) {  
4         return d.getInfo<CL_DEVICE_GLOBAL_MEM_SIZE>() >= 4_GB;  
5     });
```



Exclusive device access

- vex::Filter::Exclusive() wraps normal filters to allow exclusive access to devices.
- Useful for cluster environments.
- An alternative to NVIDIA's exclusive compute mode for other vendors hardware.
- Based on Boost.Interprocess file locks in temp directory.

```
1 vex::Context ctx( vex::Filter::Exclusive(  
2     vex::Filter::DoublePrecision &&  
3     vex::Filter::Env  
4 ) );
```

What if OpenCL context is initialized elsewhere?

- You don't *have to* initialize `vex::Context`.
- `vex::Context` is just a convenient container that holds OpenCL contexts and queues.
- `vex::Context::queue()` returns `std::vector<cl::CommandQueue>`.
This may come from *elsewhere*.

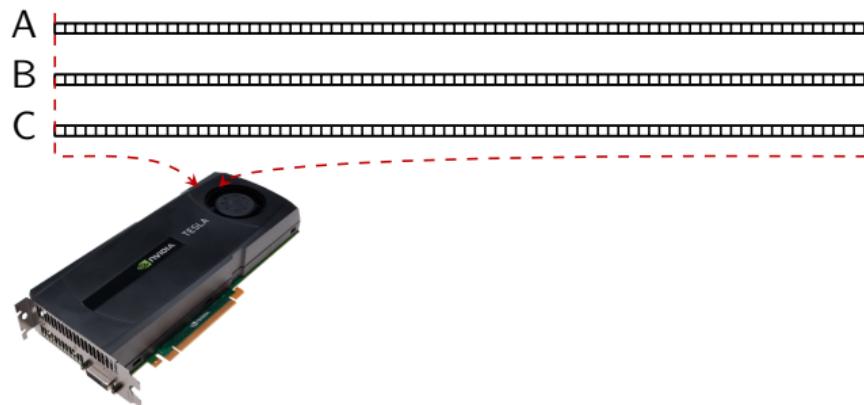
```
1 std::vector<cl::CommandQueue> my_own_vector_of_opencl_command_queues;  
2 // ...  
3 vex::vector<double> x(my_own_vector_of_opencl_command_queues, n);
```

- Each queue should correspond to a separate device.
- Different VexCL objects may be initialized with different queue lists.
- Operations are submitted to the queues of the vector that is being assigned to.

Vector allocation and arithmetic

Hello VexCL example

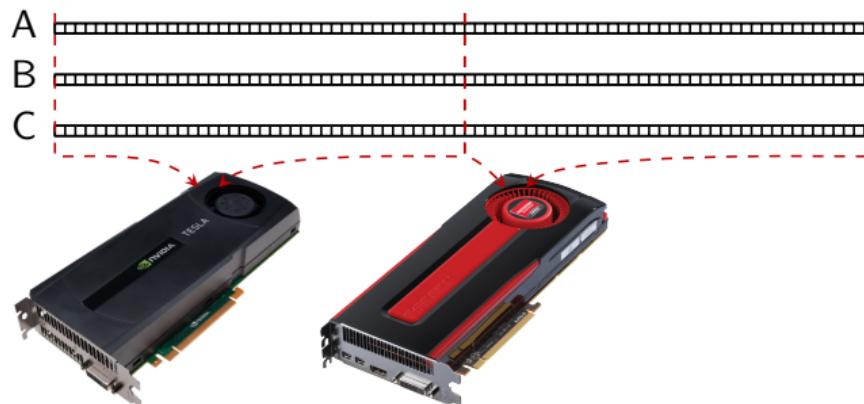
```
1 vex::Context ctx( vex::Filter::Name("Tesla") );
2
3 vex::vector<float> A(ctx.queue(), N); A = 1;
4 vex::vector<float> B(ctx.queue(), N); B = 2;
5 vex::vector<float> C(ctx.queue(), N);
6
7 C = A + B;
```



Vector allocation and arithmetic

Hello VexCL example

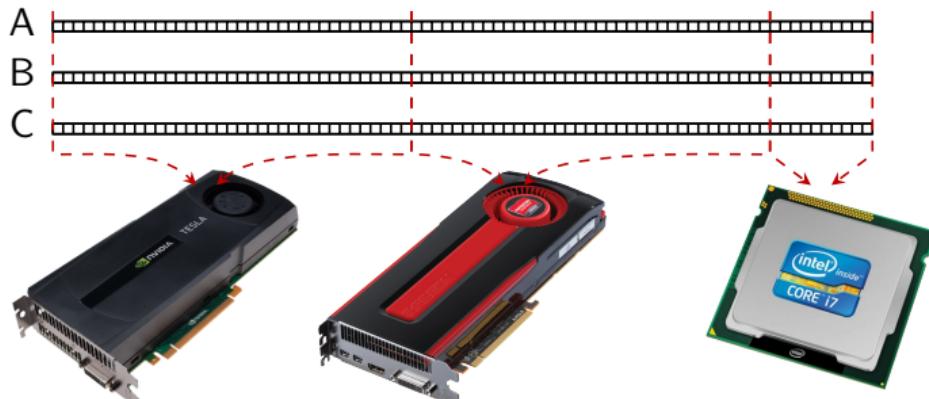
```
1 vex::Context ctx( vex::Filter::Type(CL_DEVICE_TYPE_GPU) );
2
3 vex::vector<float> A(ctx.queue(), N); A = 1;
4 vex::vector<float> B(ctx.queue(), N); B = 2;
5 vex::vector<float> C(ctx.queue(), N);
6
7 C = A + B;
```



Vector allocation and arithmetic

Hello VexCL example

```
1 vex::Context ctx( vex::Filter::DoublePrecision );  
2  
3 vex::vector<float> A(ctx.queue(), N); A = 1;  
4 vex::vector<float> B(ctx.queue(), N); B = 2;  
5 vex::vector<float> C(ctx.queue(), N);  
6  
7 C = A + B;
```



What may be used in vector expressions?

- All vectors in expression have to be *compatible*:
 - Have same size
 - Located on same devices
- What may be used:
 - Scalar values
 - Arithmetic, bitwise, logical operators
 - Builtin OpenCL functions
 - User-defined functions

```
1 std :: vector<float> x(n);
2 std :: generate(x.begin(), x.end(), rand);
3
4 vex::vector<float> X(ctx.queue(), x);
5 vex::vector<float> Y(ctx.queue(), n);
6 vex::vector<float> Z(ctx.queue(), n);
7
8 Y = 42;
9 Z = sqrt(2 * X) + pow(cos(Y), 2.0);
```

Reductions

- Class `vex::Reducer<T, kind>` allows to reduce arbitrary *vector expression* to a single value of type T.
- Supported reduction kinds: SUM, MIN, MAX

Inner product

```
1 vex::Reducer<double, vex::SUM> sum(ctx.queue());  
2 double s = sum(x * y);
```

Number of elements in x between 0 and 1

```
1 vex::Reducer<size_t, vex::SUM> sum(ctx.queue());  
2 size_t n = sum( (x > 0) && (x < 1) );
```

Maximum distance from origin

```
1 vex::Reducer<double, vex::MAX> max(ctx.queue());  
2 double d = max( sqrt(x * x + y * y) );
```

User-defined functions

- Users may define functions to be used in vector expressions:
 - Provide function body
 - Define return type and argument types

Defining a function

```
1 extern const char between_body[] = "return prm1 <= prm2 && prm2 <= prm3;";  
2 UserFunction<between_body, bool(double, double, double)> between;
```

Using a function: number of 2D points in first quadrant

```
1 size_t points_in_1q( const vex::Reductor<size_t, vex::SUM> &sum,  
2           const vex::vector<double> &x, const vex::vector<double> &y )  
3 {  
4     return sum( between(0.0, atan2(y, x), M_PI/2) );  
5 }
```

Sparse matrix – vector products

- Class `vex::SpMat<T>` holds representation of a sparse matrix on compute devices.
- Constructor accepts matrix in common CRS format (row indices, columns and values of nonzero entries).
- SpMV may only be used in additive expressions.

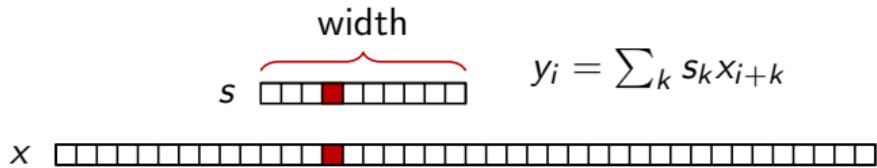
Construct matrix

```
1 vex::SpMat<double> A(ctx.queue(), n, n, row.data(), col.data(), val.data());
```

Compute residual value

```
2 // vex::vector<double> u, f, r;  
3 r = f - A * u;  
4 double res = max( fabs(r) );
```

Simple stencil convolutions



- Simple stencil is based on a 1D array, and may be used for:
 - Signal filters (e.g. averaging)
 - Differential operators with constant coefficients
 - ...

Moving average with 5-points window

```
1 std::vector<double> sdata(5, 0.2);
2 vex::stencil<double> s(ctx.queue(), sdata, 2 /* center */);
3
4 y = x * s;
```

User-defined stencil operators

- Define efficient arbitrary stencil operators:
 - Return type
 - Stencil dimensions (width and center)
 - Function body

Example: nonlinear operator

$$y_i = x_i + (x_{i-1} + x_{i+1})^3$$

Implementation

```
1 extern const char custom_op_body[] =
2     "double t = X[-1] + X[1];\n"
3     "return X[0] + t * t * t;"  

4
5 vex::StencilOperator<double, 3 /*width*/, 1 /*center*/, custom_op_body>
6     custom_op(ctx.queue());
7
8 y = custom_op(x);
```

Multivectors

- vex::multivector<T,N> holds N instances of equally sized vex::vector<T>
- Supports all operations that are defined for vex::vector<>.
- Transparently dispatches the operations to the underlying components.
- vex::multivector :: **operator**(uint k) returns k-th component.

```
1 vex::multivector<double, 2> X( ctx.queue(), N), Y( ctx.queue(), N);
2 vex::Reductor<double, vex::SUM> sum(ctx.queue());
3 vex::SpMat<double> A( ctx.queue(), ... );
4 std::array<double, 2> v;
5
6 // ...
7
8 X = sin(v * Y + 1);           //  $X(k) = \sin(v[k] * Y(k) + 1)$ ;
9 v = sum( between(0, X, Y) );   //  $v[k] = \text{sum}(\text{between}(0, X(k), Y(k)))$ ;
10 X = A * Y;                  //  $X(k) = A * Y(k)$ ;
```

Multiexpressions

- Sometimes an operation cannot be expressed with simple multivector arithmetics.

Example: rotate 2D vector by an angle

$$\begin{aligned}y_0 &= x_0 \cos \alpha - x_1 \sin \alpha, \\y_1 &= x_0 \sin \alpha + x_1 \cos \alpha.\end{aligned}$$

- Multiexpression is a tuple of normal vector expressions
- Its assignment to a multivector is functionally equivalent to componentwise assignment, but results in a single kernel launch.

Multiexpressions

- Multiexpressions may be used with multivectors:

```
1 // double alpha;
2 // vex::multivector<double,2> X, Y;
3
4 Y = std::tie(
5     X(0) * cos(alpha) - X(1) * sin(alpha),
6     X(0) * sin(alpha) + X(1) * cos(alpha) );
```

- and with tied vectors:

```
1 // vex::vector<double> alpha;
2 // vex::vector<double> oldX, oldY, newX, newY;
3
4 vex::tie(newX, newY) = std::tie(
5     oldX * cos(alpha) - oldY * sin(alpha),
6     oldX * sin(alpha) + oldY * cos(alpha) );
```

- Any expression that is assignable to a vector is valid in a multiexpression.

Copies between host and device memory

```
1 vex::vector<double> X;  
2 std::vector<double> x;  
3 double c_array[100];
```

Simple copies

```
1 vex::copy(X, x); // From device to host.  
2 vex::copy(x, X); // From host to device.
```

STL-like range copies

```
1 vex::copy(X.begin(), X.end(), x.begin());  
2 vex::copy(X.begin(), X.begin() + 100, x.begin());  
3 vex::copy(c_array, c_array + 100, X.begin());
```

Inspect or set single element (*slow*)

```
1 vex::copy(X, x);  
2 assert(x[42] == X[42]);  
3 X[0] = 0;
```

Motivation
oooooooooo

Basic interface
oooooooooooooo

Kernel builder
oooooooooo

Performance
oo

Implementation details
oooo

Conclusion

1 Motivation

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Converting generic C++ algorithms to OpenCL kernels*

*Restrictions applied

Motivating example

- Let's solve an ODE!
- Let's do it with Boost.odeint!

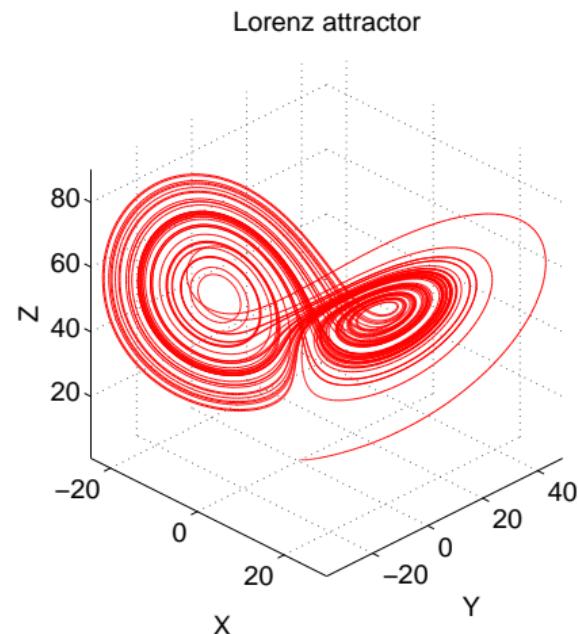
- Lorenz attractor system:

$$\dot{x} = -\sigma(x - y),$$

$$\dot{y} = Rx - y - xz,$$

$$\dot{z} = -bz + xy.$$

- We want to solve large number of Lorenz systems, each for a different value of R .



odeint setup

1. System functor

```
1 typedef vex::vector<double> vector_type;
2 typedef vex::multivector<double, 3> state_type;
3
4 struct lorenz_system {
5     const vector_type &R;
6     lorenz_system(const vector_type &R ) : R(R) { }
7
8     void operator()(const state_type &x, state_type &dxdt, double t) {
9         dxdt = std::tie(
10             sigma * ( x(1) - x(0) ),
11             R * x(0) - x(1) - x(0) * x(2),
12             x(0) * x(1) - b * x(2)
13         );
14     }
15 };
```

odeint setup

2. Integration

```
1 state_type X( ctx.queue(), n );
2 vector_type R( ctx.queue(), r );
3
4 // ... initialize X and R here ...
5
6 odeint::runge_kutta4<
7     state_type, double, state_type, double,
8     odeint::vector_space_algebra, odeint::default_operations
9 > stepper;
10
11 odeint::integrate_const(stepper, lorenz_system(R), X, 0.0, t_max, dt);
```

- That was easy!

odeint setup

2. Integration

```
1 state_type X( ctx.queue(), n );
2 vector_type R( ctx.queue(), r );
3
4 // ... initialize X and R here ...
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- That was easy! And fast!

odeint setup

2. Integration

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- That was easy! And fast! **But,**

odeint setup

2. Integration

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7     state_type, double, state_type, double,
8     odeint::vector_space_algebra, odeint::default_operations
9 > stepper;
10
11 odeint::integrate_const(stepper, lorenz_system(R), X, 0.0, t_max, dt);
```

- That was easy! And fast! But,
 - Runge-Kutta method uses 4 temporary state variables (here stored on GPU).
 - Single Runge-Kutta step results in several kernel launches.

What if we did this manually?

- Create single monolithic kernel that does one step of Runge-Kutta method.
- Launch the kernel in a loop.
- This is ≈ 10 times faster!

```
1 double3 lorenz_system(double r, double sigma, double b, double3 s) {
2     return (double3)(
3         sigma * (s.y - s.x),
4         r * s.x - s.y - s.x * s.z,
5         s.x * s.y - b * s.z
6     );
7 }
8
9 kernel void lorenz_ensemble(
10    ulong n, double sigma, double b,
11    const global double *R,
12    global double *X,
13    global double *Y,
14    global double *Z
15 )
16 {
17     double r;
18     double3 s, dsdt, k1, k2, k3, k4;
19
20     for( size_t gid = get_global_id(0); gid < n; gid += get_global_size(0) ) {
21         r = R[gid];
22         s = (double3)(X[gid], Y[gid], Z[gid]);
23
24         k1 = dt * lorenz_system(r, sigma, b, s);
25         k2 = dt * lorenz_system(r, sigma, b, s + 0.5 * k1);
26         k3 = dt * lorenz_system(r, sigma, b, s + 0.5 * k2);
27         k4 = dt * lorenz_system(r, sigma, b, s + k3);
28
29         s += (k1 + 2 * k2 + 2 * k3 + k4) / 6;
30
31         X[gid] = s.x; Y[gid] = s.y; Z[gid] = s.z;
32     }
33 }
```

What if we did this manually?

- Create single monolithic kernel that does one step of Runge-Kutta method.
- Launch the kernel in a loop.
- This is ≈ 10 times faster! But,

```
1 double3 lorenz_system(double r, double sigma, double b, double3 s) {
2     return (double3(
3         sigma * (s.y - s.x),
4         r * s.x - s.y - s.x * s.z,
5         s.x * s.y - b * s.z
6     ));
7 }
8
9 kernel void lorenz_ensemble(
10    ulong n, double sigma, double b,
11    const global double *R,
12    global double *X,
13    global double *Y,
14    global double *Z
15 )
16 {
17     double r;
18     double3 s, dsdt, k1, k2, k3, k4;
19
20     for( size_t gid = get_global_id(0); gid < n; gid += get_global_size(0) ) {
21         r = R[gid];
22         s = (double3)(X[gid], Y[gid], Z[gid]);
23
24         k1 = dt * lorenz_system(r, sigma, b, s);
25         k2 = dt * lorenz_system(r, sigma, b, s + 0.5 * k1);
26         k3 = dt * lorenz_system(r, sigma, b, s + 0.5 * k2);
27         k4 = dt * lorenz_system(r, sigma, b, s + k3);
28
29         s += (k1 + 2 * k2 + 2 * k3 + k4) / 6;
30
31         X[gid] = s.x; Y[gid] = s.y; Z[gid] = s.z;
32     }
33 }
```

What if we did this manually?

- Create single monolithic kernel that does one step of Runge-Kutta method.
- Launch the kernel in a loop.
- This is ≈ 10 times faster! But,
- We lost the generality odeint offers!

```
1 double3 lorenz_system(double r, double sigma, double b, double3 s) {
2     return (double3(
3         sigma * (s.y - s.x),
4         r * s.x - s.y - s.x * s.z,
5         s.x * s.y - b * s.z
6     ));
7 }
8
9 kernel void lorenz_ensemble(
10    ulong n, double sigma, double b,
11    const global double *R,
12    global double *X,
13    global double *Y,
14    global double *Z
15 )
16 {
17     double r;
18     double3 s, dsdt, k1, k2, k3, k4;
19
20     for( size_t gid = get_global_id(0); gid < n; gid += get_global_size(0) ) {
21         r = R[gid];
22         s = (double3)(X[gid], Y[gid], Z[gid]);
23
24         k1 = dt * lorenz_system(r, sigma, b, s);
25         k2 = dt * lorenz_system(r, sigma, b, s + 0.5 * k1);
26         k3 = dt * lorenz_system(r, sigma, b, s + 0.5 * k2);
27         k4 = dt * lorenz_system(r, sigma, b, s + k3);
28
29         s += (k1 + 2 * k2 + 2 * k3 + k4) / 6;
30
31         X[gid] = s.x; Y[gid] = s.y; Z[gid] = s.z;
32     }
33 }
```

Convert generic C++ algorithms to OpenCL kernels

- ① Capture the sequence of arithmetic expressions of an algorithm.
- ② Construct OpenCL kernel from the captured sequence.
- ③ ???
- ④ Use the kernel!

Convert generic C++ algorithms to OpenCL kernels

1. Declare functor operating on vex::generator::symbolic<> values

```
1 typedef vex::generator::symbolic< double > sym_vector;
2 typedef std::array<sym_vector, 3> sym_state;
3
4 struct lorenz_system {
5     const sym_vector &R;
6     lorenz_system(const sym_vector &R) : R(R) {}
7     void operator()(const sym_state &x, sym_state &dxdt, double t) const {
8         dxdt[0] = sigma * (x[1] - x[0]);
9         dxdt[1] = R * x[0] - x[1] - x[0] * x [2];
10        dxdt[2] = x[0] * x[1] - b * x[2];
11    }
12};
```

Convert generic C++ algorithms to OpenCL kernels

2. Record one step of Runge-Kutta method

```
1 std :: ostringstream lorenz_body;
2 vex::generator :: set_recorder (lorenz_body);
3
4 sym_state sym_S = {{}
5     sym_vector::VectorParameter,
6     sym_vector::VectorParameter,
7     sym_vector::VectorParameter }};
8 sym_vector sym_R(sym_vector::VectorParameter, sym_vector::Const);
9
10 odeint :: runge_kutta4<
11     sym_state, double, sym_state, double,
12     odeint :: range_algebra, odeint :: default_operations
13     > stepper;
14
15 lorenz_system sys(sym_R);
16 stepper.do_step(std :: ref(sys), sym_S, 0, dt);
```

Convert generic C++ algorithms to OpenCL kernels

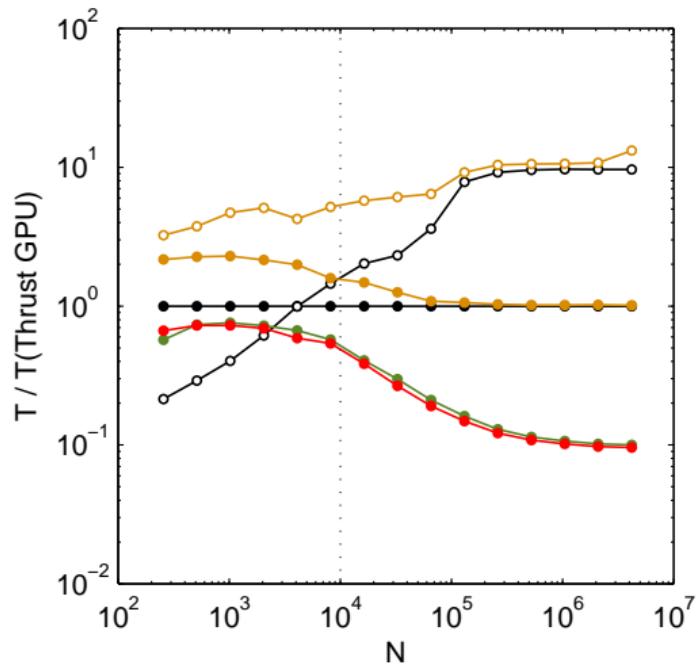
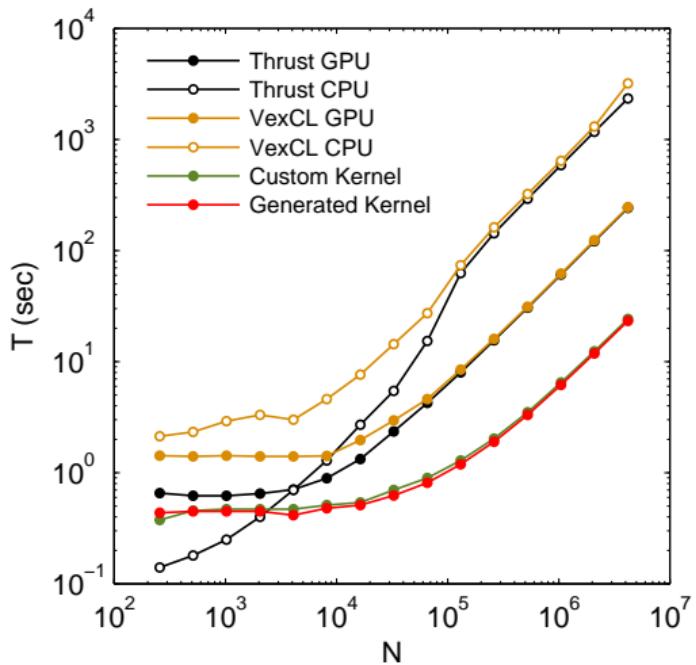
3. Generate and use OpenCL kernel

```
1 auto lorenz_kernel = vex::generator::build_kernel(ctx.queue(), "lorenz", lorenz_body.str(),
2     sym_S[0], sym_S[1], sym_S[2], sym_R);
3
4 vex::vector<double> X(ctx.queue(), n);
5 vex::vector<double> Y(ctx.queue(), n);
6 vex::vector<double> Z(ctx.queue(), n);
7 vex::vector<double> R(ctx.queue(), r);
8
9 // ... initialize X, Y, Z, and R here ...
10
11 for(double t = 0; t < t_max; t += dt) lorenz_kernel(X, Y, Z, R);
```

The restrictions

- Algorithms have to be embarrassingly parallel.
- Only linear flow is allowed (no conditionals or data-dependent loops).
- Some precision may be lost when converting constants to strings.
- Probably some other corner cases. . .

The performance results

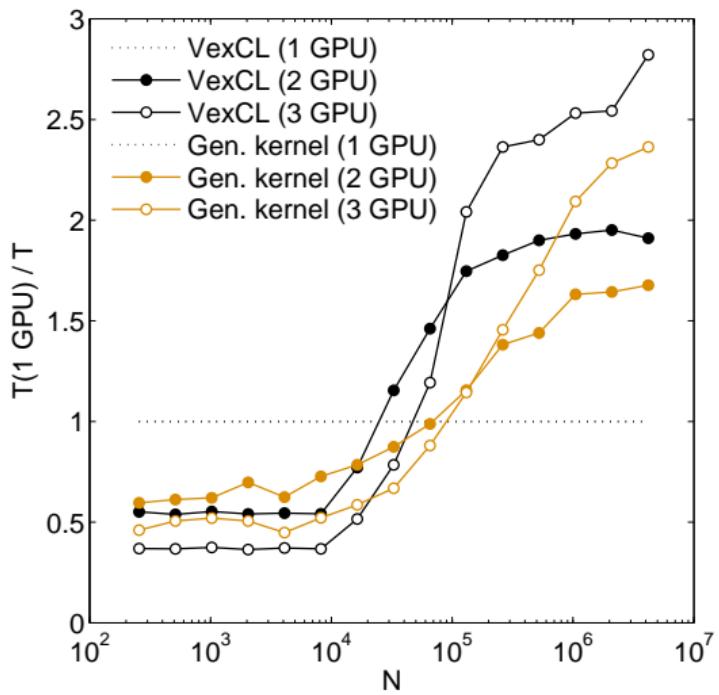


GPU: NVIDIA Tesla C2070

CPU: Intel Core i7 930

Multigpu scalability

- *Larger* problems may be solved on the same system.
- Large problems may be solved *faster*.



Motivation
oooooooooo

Basic interface
oooooooooooooo

Kernel builder
oooooooooo

Performance
oo

Implementation details
oooo

Conclusion

1 Motivation

2 Basic interface

3 Kernel builder

4 Performance

5 Implementation details

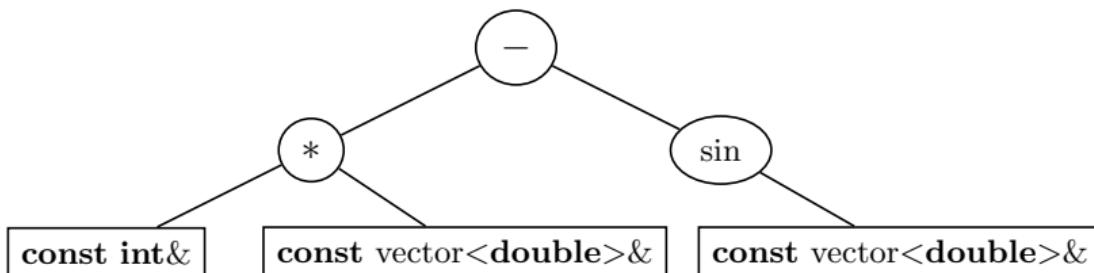
6 Conclusion

Expression trees

- VexCL is an *expression template* library
- Each expression in the code results in an expression tree evaluated at time of assignment.
 - No temporaries are created
 - Single kernel is generated and executed

Example expression

```
1 x = 2 * y - sin(z);
```

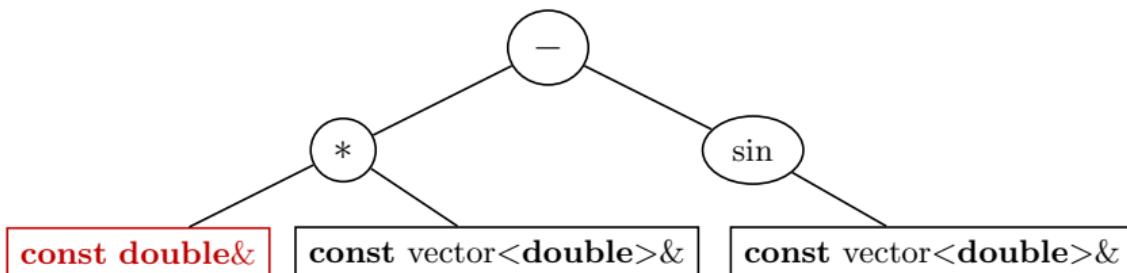


Expression trees

- VexCL is an *expression template* library
- Each expression in the code results in an expression tree evaluated at time of assignment.
 - No temporaries are created
 - Single kernel is generated and executed

Example expression

```
1 x = 2.0 * y - sin(z);
```



Kernel generation

The expression

```
1 x = 2 * y - sin(z);
```

Define VEXCL_SHOW KERNELS to see the generated code.

... results in this kernel:

```
1 kernel void minus_multiplies_term_term_sin_term(
2     ulong n,
3     global double *res,
4     int prm_1,
5     global double *prm_2,
6     global double *prm_3
7 )
8 {
9     for( size_t idx = get_global_id(0); idx < n; idx += get_global_size(0)) {
10         res[idx] = ( ( prm_1 * prm_2[idx] ) - sin( prm_3[idx] ) );
11     }
12 }
```

Performance tip

- No way to tell if two terminals refer to the same data!
- Example: finding number of points in 1st quadrant

Naive

```
1 return sum( 0.0 <= atan2(y, x) && atan2(y, x) <= M_PI/2 );
```

- x and y are read twice
- atan2 is computed twice

Using custom function

```
1 return sum( between(0.0, atan2(y, x), M_PI/2) );
```

Custom kernels

It is possible to use custom kernels with VexCL vectors

```
1 vex::vector<float> x(ctx.queue(), n);
2
3 for(uint d = 0; d < ctx.size(); d++) {
4     cl::Program program = build_sources(ctx.context(d),
5         "kernel void dummy(ulong size, global float *x) {\n"
6         "    x[get_global_id(0)] = 4.2;\n"
7         "}\n");
8
9     cl::Kernel dummy(program, "dummy");
10
11    dummy.setArg(0, static_cast<cl_ulong>(x.part_size(d)));
12    dummy.setArg(1, x(d));
13
14    ctx.queue(d).enqueueNDRangeKernel(dummy, cl::NullRange, x.part_size(d), cl::NullRange);
15 }
```

Conclusion and Questions

- VexCL allows to write compact and readable code without sacrificing too much performance.
- Multiple compute devices are employed transparently.
- Supported compilers (don't forget to enable C++11 features):
 - GCC v4.6
 - Clang v3.1
 - MS Visual C++ 2010 (partially)
- <https://github.com/demidov/vexcl>



Conjugate gradients method

Solve linear equations system $Au = f$

```
1 void cg::solve(const vex::SpMat<double> &A, const vex::vector<double> &f, vex::vector<double> &u) {
2     // Member fields:
3     // vex::vector<double> r, p, q;
4     // Reductor<double,MAX> max; Reductor<double,SUM> sum;
5
6     double rho1 = 0, rho2 = 1;
7     r = f - A * u;
8
9     for(int iter = 0; max( fabs(r) ) > 1e-8 && iter < n; iter++) {
10         rho1 = sum(r * r);
11
12         if (iter == 0) {
13             p = r;
14         } else {
15             double beta = rho1 / rho2;
16             p = r + beta * p;
17         }
18
19         q = A * p;
20
21         double alpha = rho1 / sum(p * q);
22
23         u += alpha * p;
24         r -= alpha * q;
25
26         rho2 = rho1;
27     }
28 }
```

The generated kernel (is ugly)

```

1 kernel void lorenz(
2 ulong n,
3 global double* p_var0,
4 global double* p_var1,
5 global double* p_var2,
6 global const double* p_var3
7 )
8 {
9     size_t idx = get_global_id(0);
10    if (idx < n) {
11        double var0 = p_var0[idx];
12        double var1 = p_var1[idx];
13        double var2 = p_var2[idx];
14        double var3 = p_var3[idx];
15        double var4;
16        double var5;
17        double var6;
18        double var7;
19        double var8;
20        double var9;
21        double var10;
22        double var11;
23        double var12;
24        double var13;
25        double var14;
26        double var15;
27        double var16;
28        double var17;
29        double var18;
30        var4 = (1.00000000000e+01 * (var1 - var0));
31        var5 = ((var3 * var0) - var1) - (var0 * var2);
32        var6 = ((var0 * var1) - (2.66666666666e-00 * var2));
33        var7 = ((1.00000000000e+00 * var0) + (5.00000000000e-03 * var4));
34        var8 = ((1.00000000000e+00 * var1) + (5.00000000000e-03 * var5));
35        var9 = ((1.00000000000e+00 * var2) + (5.00000000000e-03 * var6));
36        var10 = (1.00000000000e+01 * (var8 - var7));
37        var11 = ((var1 * var7) - var8) - (var7 * var9);
38        var12 = (var7 * var8) - (2.66666666666e+00 * var9);
39        var7 = (((1.00000000000e+00 * var0) + (0.00000000000e+00 * var4)) + (5.00000000000e-03 * var10));
40        var8 = (((1.00000000000e+00 * var1) + (0.00000000000e+00 * var5)) + (5.00000000000e-03 * var11));
41        var9 = (((1.00000000000e+00 * var2) + (0.00000000000e+00 * var6)) + (5.00000000000e-03 * var12));
42        var13 = (1.00000000000e+01 * (var8 - var7));
43        var14 = ((var1 * var7) - var8) - (var7 * var9);
44        var15 = (var7 * var8) - (2.66666666666e+00 * var9);
45        var7 = (((1.00000000000e+00 * var0) + (0.00000000000e+00 * var4)) + (0.00000000000e+00 * var10)) + (1.00000000000e-02 * var13));
46        var8 = (((1.00000000000e+00 * var1) + (0.00000000000e+00 * var5)) + (0.00000000000e+00 * var11)) + (1.00000000000e-02 * var14);
47        var9 = (((1.00000000000e+00 * var2) + (0.00000000000e+00 * var6)) + (0.00000000000e+00 * var12)) + (1.00000000000e-02 * var15));
48        var16 = (1.00000000000e+01 * (var8 - var7));
49        var17 = ((var1 * var7) - var8) - (var7 * var9);
50        var18 = (var7 * var8) - (2.66666666666e+00 * var9);
51        var7 = (((1.00000000000e+00 * var0) + (1.66666666666e-03 * var4)) + (3.33333333333e-03 * var10)) + (3.33333333333e-03 * var13)) + (1.66666666666e-03 * var16));
52        var1 = (((1.00000000000e+00 * var1) + (1.66666666666e-03 * var5)) + (3.33333333333e-03 * var11)) + (3.33333333333e-03 * var14)) + (1.66666666666e-03 * var17));
53        var2 = (((1.00000000000e+00 * var2) + (1.66666666666e-03 * var6)) + (3.33333333333e-03 * var12)) + (3.33333333333e-03 * var15)) + (1.66666666666e-03 * var18));
54        p_var0[idx] = var0;
55        p_var1[idx] = var1;
56        p_var2[idx] = var2;
57    }
58 }
```