

# Manifold Actors: Extending the C++ Actor Framework to Heterogeneous Many-Core Machines using OpenCL

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# Motivation

- Many-core devices available for general-purpose computation
  - Specialized components can vastly outperform CPUs
  - Available on servers, desktops and mobile devices
- Require specialized frameworks and programming models
  - Often low-level APIs
  - Management overhead
  - Specialized syntax or language
  
- Can we use the actor model to abstract over heterogenous computing?

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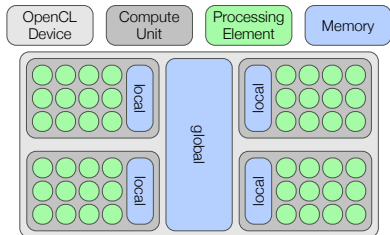
# Outline

- 1 Heterogeneous Computing
- 2 The C++ Actor Framework (CAF)
- 3 The OpenCL Actor
- 4 Performance Evaluation
- 5 Conclusion & Outlook

# OpenCL


## Our Starting Point

- Vendor-independent standard
- Maintained by the Khronos Group
- Supports a wide range of hardware
  - GPUs, CPUs, Accelerators
  - Simplified device model
- Concurrent execution of “kernels”
  - Executed in 3D index space
  - Command queues for device interaction
  - Offers asynchronous API



The OpenCL device model.

## The C++ Actor Framework

- Actor library written in C++11
  - Low memory footprint
  - Fast, lock-free mailbox implementation
  - Work-stealing scheduler
  - Type-safe message passing
- Focus on scalability
  - Up to multi-core machines
  - Down to embedded devices with 
- Working on Runtime Inspection and Configuration tools
  - Ease debugging of distributed systems
  - View on messages and connections in the system
  - Interact by injecting messages

# The OpenCL Actor

- Intra-actor concurrency
  - Map the OpenCL workflow to actor message passing
  - Many kernel instances executed in parallel
  - Kernels may run in parallel on the device (if supported)
- Hide complexity
  - OpenCL provides a low-level interface
  - Similar steps in each program
  - Management of OpenCL setup and device initialization
- Integrate seamlessly
  - Use the same handle types as other actors
  - Hide physical deployment
  - Network transparency, monitoring and error propagation

# Use Case

```
constexpr const char* source = R"__(
__kernel void m_mult(__global float* matrix1,
                    __global float* matrix2,
                    __global float* output) {
    size_t size = get_global_size(0);
    size_t x = get_global_id(0);
    size_t y = get_global_id(1);
    float result = 0;
    for (size_t idx = 0; idx < size; ++idx) {
        result += matrix1[idx + y * size]
                * matrix2[x + idx * size];
    }
    output[x + y * size] = result;
})__";
constexpr const char* name = "m_mult";
```



# Use Case

```
constexpr const char* source = R"__(
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```

Marks function as an OpenCL kernel,  
which always returns `void`.

# Use Case

```
constexpr const char* source = R"__(
__kernel void m_mult(__global float* matrix1,
__global float* matrix2,
__global float* output) {
    size_t size = get_global_size(0);
    size_t x = get_global_id(0);
    size_t y = get_global_id(1);
    float result = 0;
    for (int idx = 0; idx < size; ++idx) {
        result += matrix1[idx + y * size]
            * matrix2[x + idx * size];
    }
    output[x + y * size] = result;
})__";
constexpr const char* name = "m_mult";
```

Specifies memory regions of arguments.

# Use Case

```
constexpr const char* source = R"__(
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}__";
constexpr const char* name = "m_mult";
```

Identifies work-items in the defined index space.

# Use Case

```
constexpr const char* source = R"__(
__kernel void m_mult(__global float* matrix1,
                    __global float* matrix2,
                    __global float* output) {
    size_t size = get_global_size(0);
    size_t x = get_global_id(0);
    size_t y =
float result;
for (size_t
Data-parallel assignment
of results.
        ++idx) {
    result += matrix1[idx + y * size]
        * matrix2[x + idx * size];
}
output[x + y * size] = result;
})__";
constexpr const char* name = "m_mult";
```

# Interface

```
int main() {
    using fvec = std::vector<float>;
    constexpr size_t mx_dim = 1024;
    auto worker = spawn_cl(
        source, name,
        spawn_config{dim_vec{mx_dim, mx_dim}},
        in<fvec>{}, in<fvec>{}, out<fvec>{}
    );
    auto m = create_matrix(mx_dim * mx_dim);
    scoped_actor self;
    self->sync_send(worker, m, m).await(
        [](const fvec& result) {
            print_as_matrix(result);
        }
    );
}
```

# Interface

```
int main() {  
    using fvec = std::vector<float>;  
    constexpr size_t mx_dim = 1024;  
    auto worker = spawn_cl(  
        source, name,  
        spawn_config{dim_vec{mx_dim, mx_dim}},  
        in<fvec>{}, in<fvec>{}, out<fvec>{}  
    );  
    auto scope = ...; // ...  
    self->sync_send(worker, m, m).await(  
        [](const fvec& result) {  
            print_as_matrix(result);  
        }  
    );  
}
```

Function to create an OpenCL actor.  
Requires OpenCL specific parameters

# Interface

```
int main() {  
    using fvec = std::vector<float>;  
    constexpr size_t mx_dim = 1024;  
    auto worker = spawn_cl(  
        source, name,  
        spawn_config{dim_vec{mx_dim, mx_dim}},  
        in<fvec>{}, in<fvec>{}, out<fvec>{}  
    );  
    auto m = Code and name of the kernel. im);  
    scoped_await(worker, m);  
    self->sync_send(worker, m, m).await(  
        [](const fvec& result) {  
            print_as_matrix(result);  
        }  
    );  
}
```

# Interface

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int main() {  
    using fvec = std::vector<float>;  
    constexpr size_t mx_dim = 1024;  
    auto worker = spawn_cl(  
        source, name,  
        spawn_config{dim_vec{mx_dim, mx_dim}},  
        in<fvec>{}, in<fvec>{}, out<fvec>{}  
    );  
    auto m = create_matrix(mx_dim * mx_dim);  
    ...  
};
```

Specifies index space for the execution.  
Accepts more parameters, .e.g., for the compiler.



# Interface

```
int main() {  
    using fvec = std::vector<float>;  
    constexpr size_t mx_dim = 1024;  
    auto worker = spawn_cl(  
        source, name,  
        spawn config{dim vec{mx dim, mx dim}},  
        in<fvec>{}, in<fvec>{}, out<fvec>{}  
    );  
    auto m = create matrix(mx dim * mx_dim);  
    await(  
        print_as_matrix(result),  
    );  
}
```

Specifies the signature of the kernel,  
i.e., argument type, position and if it  
is input, output or both.

# Interface

```
int main() {  
    using fvec = std::vector<float>;  
    constexpr size_t mx_dim = 1024;  
    auto worker = spawn_cl(  
        source { Sends a message to the actor and  
                prints the result. },  
        spawn { m }},  
    in<fvec> {}, in<fvec> {}, out<fvec> {}  
);  
    auto m = create_matrix(mx_dim * mx_dim);  
    scoped_actor self;  
    self->sync_send(worker, m, m).await(  
        [] (const fvec& result) {  
            print_as_matrix(result);  
        }  
    );  
}
```

# Limitations

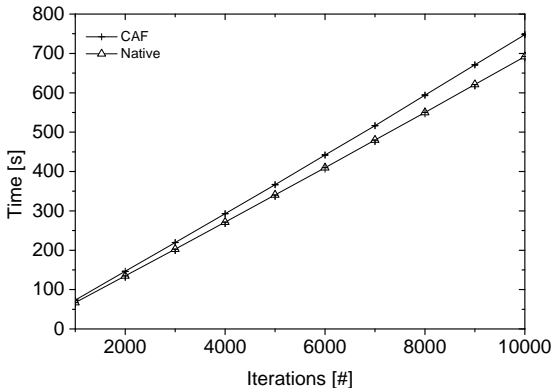
- Current limitations of OpenCL actors
  - No message passing from OpenCL context
  - Actor creation only from the CPU context
  - Behavior cannot be changed at runtime
  - OpenCL actors have no state
- Depends on available hardware & drivers

# Baseline Benchmark

- Performance comparison to native OpenCL
  - Initialization only occurs once
  - Management performed in OpenCL callbacks
- Multiply two  $1000 \times 1000$  matrices
  - Chain of independent calculations
  - Same code for kernel
  - No simultaneous executions
- Environment
  - Desktop with OSX (10.11)
  - NVIDIA GeForce GTX 780M GPU, OpenCL Version 1.2
  - Intel Core i7 clocked at 3.5 GHz

# Results

## Baseline Benchmark



- OpenCL runtime exhibits a smaller slope
- Indication of a consistent overhead for message passing
- Not reachable with a realistic application

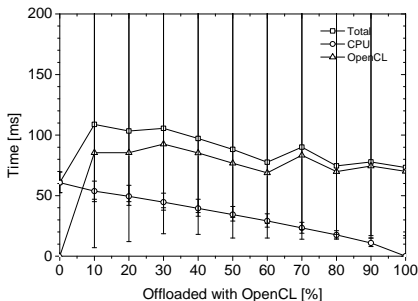
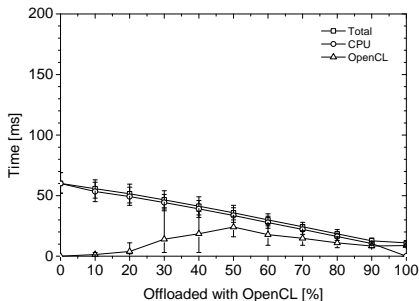
# Scaling Benchmark

- Scalability in a heterogeneous scenario
  - Examine relation to problem size
  - Take different hardware into account
- Calculation of the Mandelbrot set
  - Easily dividable into independent tasks
  - Inner cut with a balanced processing complexity
  - Offload in steps of 10%, from 0% to 100%
- Environment
  - Server with Linux (kernel 3.19)
  - Nvidia Tesla C2075 GPU, OpenCL Version 1.1
  - Intel Xeon Phi 5110P, OpenCL Version 1.2
  - Two twelve-core Intel Xeon CPUs clocked at 2.5 GHz

# Results: Small Problem

## Scaling Benchmark

Small workload (1920 × 1080 pixels, 100 iterations) offloaded to a Tesla GPU (left) and a Xeon Phi accelerator (right).

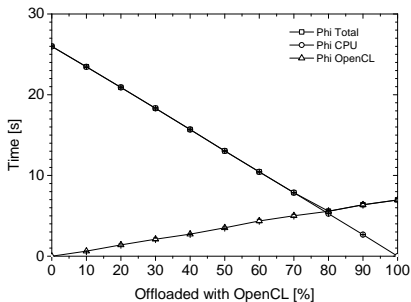
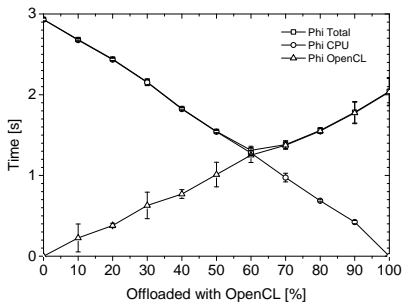


- The Tesla exhibits excellent scaling behavior
- Calculating 10% on the CPU takes longer than 100% on the Tesla
- Xeon Phi shows large overhead

# Results: Large Problems

## Scaling Benchmark

Large workload ( $16\,000 \times 16\,000$  pixels) on the Phi for 100 (left) and 1000 iterations (right).



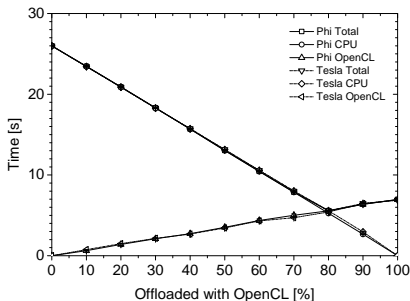
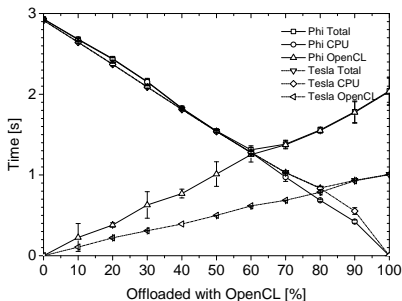
- Best performance no longer reached by offloading everything



# Results: Large Problems

## Scaling Benchmark

Large workload ( $16\,000 \times 16\,000$  pixels) on the Tesla and Phi for 100 (left) and 1000 iterations (right).



- Best performance no longer reached by offloading everything
- Performance of Phi and Tesla converge

# Conclusion & Outlook

- Introducing OpenCL actors
  - Handles most management tasks
  - Small overhead compared to native OpenCL
  - Good scalability when offloading work
  - Further benchmarks in the paper
- A native implementation of the actor model: CAF
  - High level of abstraction without sacrificing performance
  - Small memory footprint & efficient program execution
  - Freely available & open source (BSD or Boost)
- Future directions
  - Improve communication between OpenCL actors
  - Explore how OpenCL actors can hold state
  - Load-balancing across local and remote devices



Thank you for your attention.  
Questions?

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Developer blog: <http://actor-framework.org>  
Sources: <https://github.com/actor-framework/>  
iNET: <https://inet.cpt.haw-hamburg.de>