

# DtCraft: A Distributed Execution Engine for Compute-intensive Applications

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*Boost your productivity in writing parallel code!*

# Agenda

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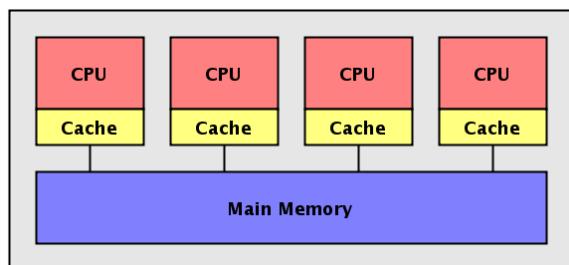


- Express your parallelism in the right way
- Boost your productivity in writing parallel code
- Leverage your time to produce promising results

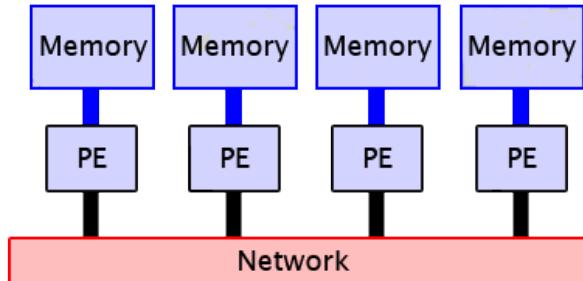
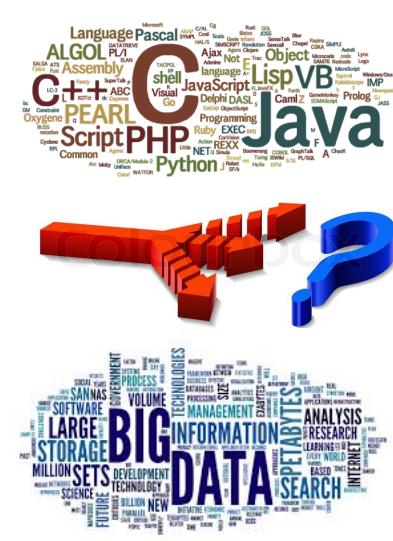
# Motivation



- “*We need a new parallel/distributed timing analysis method to deal with the large design complexities,*”  
**IBM Timing Group, Fishkill, NY, 2015**
  - Explore a feasible framework
  - Prototype a distributed timer
  - Scale to billions of transistors



## Multi-threaded timer

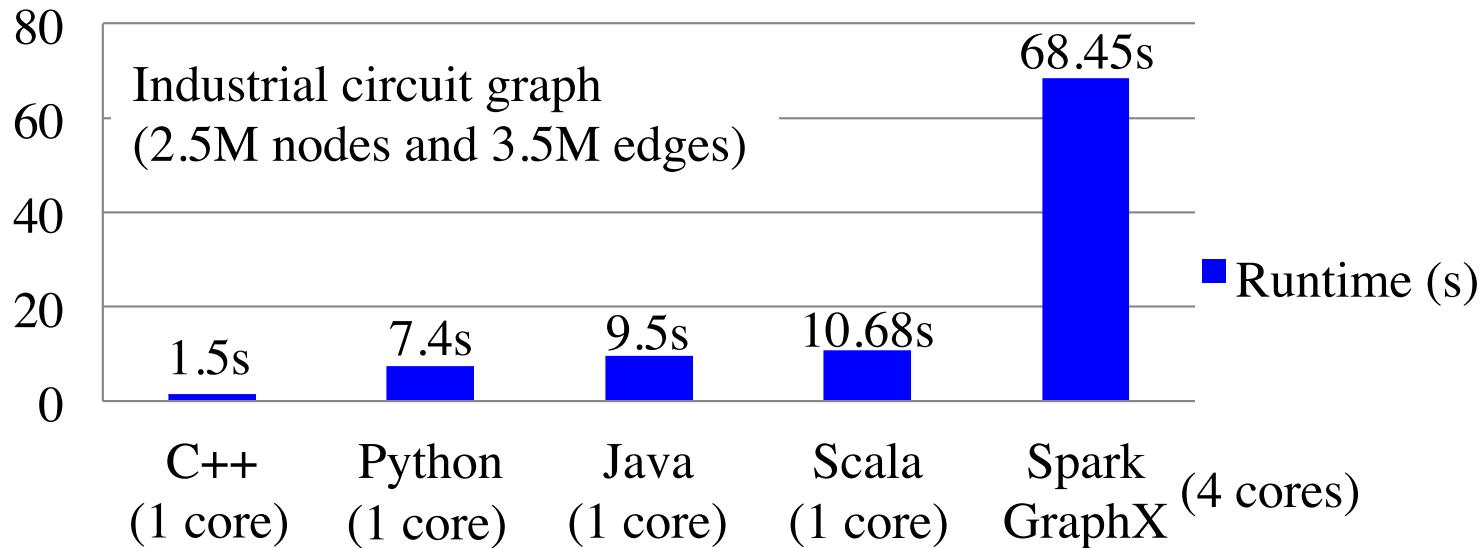


## Distributed timer

# Big data is NOT an easy fit in EDA!



## Runtime comparison on arrival time propagation



Method	Spark 1.4 (RDD + GraphX Pregel)	Java (SP)	C++ (SP)
Runtime (s)	68.45	9.5	<b>1.50</b>

# A “hard-coded” distributed timer



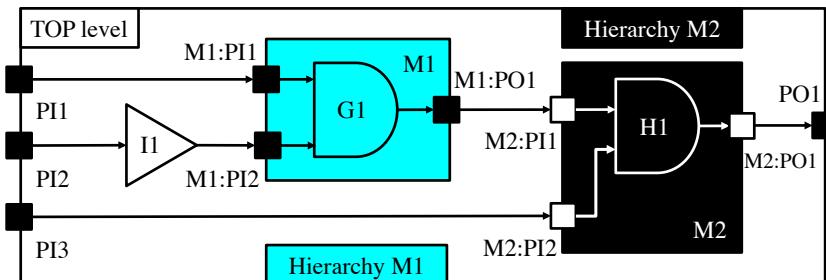
## □ General design partitions

- Logical, physical, or hierarchical partitions
- Design data are stored in a shared storage (e.g., NFS, GPFS)

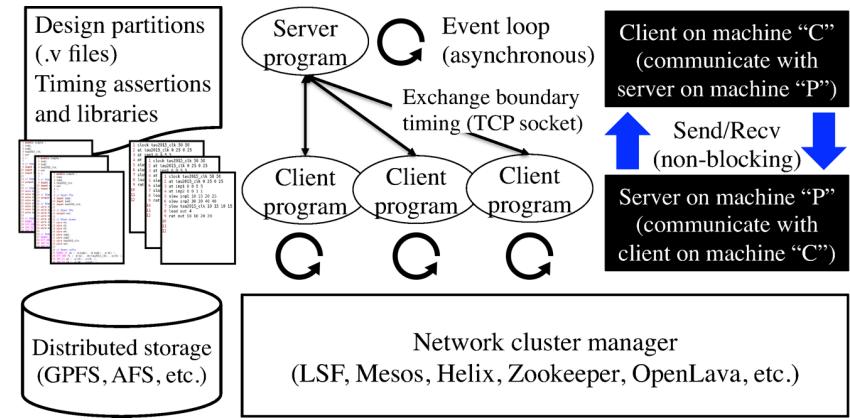
## □ Single-server multiple-client model

- Server is the centralized coordinator
- Clients exchange boundary timing with server

- Non-blocking IO
- Event-driven programming
- Serialization/Deserialization



Three partitions, top-level, M1, and M2  
(given by design teams)



# Observations

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- **Big-data is not an easy fit in EDA**
  - IO-bound vs CPU-bound
  - Unstructured vs Structured
  - JVM vs C/C++
- **Hard-coded method is error-prone and not scalable**
  - Expose to the low-level socket message passing
  - Move data between compute nodes' memories
  - Manage the cluster resource by yourselves
  - Difficult to maintain between software generations
  - Cause you a significant amount of coding efforts
- **Want parallel programming at scale far more *productive***
  - Better productivity means better performance for most people

# What does “Productivity” mean to you?

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## Programming language

- “I use Python/Matlab/Scala to prototype my project”

## Transparency

- “I use Hadoop/Spark to express my parallel computations without understanding architecture-specific details”

## Performance

- “I use C/C++/Fortran/MPI to ensure full control over resources to achieve the best CPU and memory performance”

## DtCraft project

- “We let *less-experienced* users express their parallel computing workload without taking away the control over system details to achieve *high performance*, using our *groovy* API written in modern C++17”

# Why is being “Productive” important?



## □ Code costs are more than machine costs

- Hardware is a commodity resource
- Coding takes people and time



## □ I hate writing boilerplate code

- Redundant steps to write parallel code

```
// C++ thread example
std::thread t1([](){ /* do something */ });
std::thread t2([](){ /* do another thing */ });
t1.join(); // release thread 1 resource
t2.join(); // release thread 2 resource
```

2016 average software engineer  
salary > 100K USD

- Code becomes messy when data dependencies exist

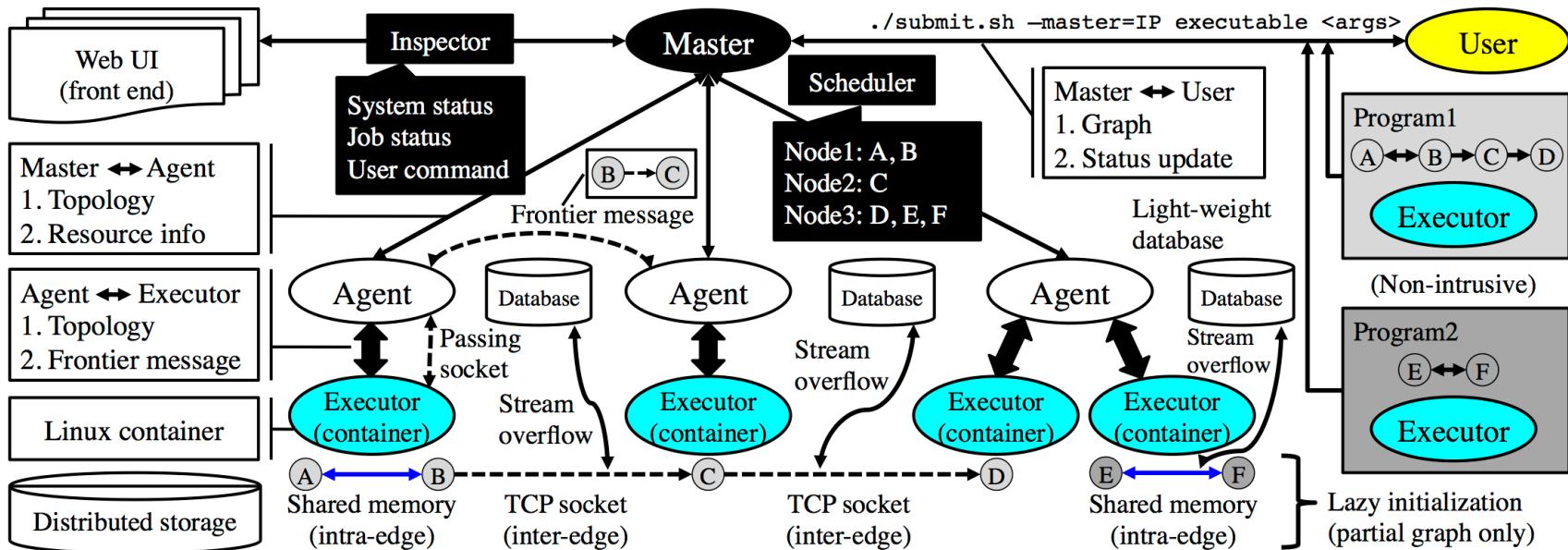
## □ We want computationally productive code

- The cloud businesses reduce the hardware factor
- Everything must be parallel moving forward

# DtCraft – A distributed execution engine



- Modernize yourself with C++17
- Express your workload in our groovy API
- Stay away from difficult concurrency controls
- Make the most use of cluster resources
- Gain security and reliability with Linux container



# Stream graph programming model



```
class Stream {  
    weak_ptr<OutputStream> ostream();  
    weak_ptr<InputStream> istream();  
    function<Signal(Vertex&, OutputStream&)> on_ostream;  
    function<Signal(Vertex&, InputStream&)> on_istream;  
};  
  
class Vertex {  
    shared_ptr<OutputStream> ostream(key_type) const;  
    shared_ptr<InputStream> istream(key_type) const;  
    function<void(Vertex&)> on;  
    unordered_map<key_type, Stream*> streams;  
};  
  
class Graph {  
    VertexDescriptor vertex();  
    StreamDescriptor stream(key_type, key_type);  
    ContainerDescriptor container();  
};  
  
class Executor : public Reactor {  
    Executor(Graph&);  
};
```

*Only a single executable is required to enable distributed execution!*

## □ Graph

- Vertex and stream creation
- Resource assignment

## □ Vertex

- One-time callback
- Access adjacent streams

## □ Stream

- Level-triggered I/O callback
- Close stream on return

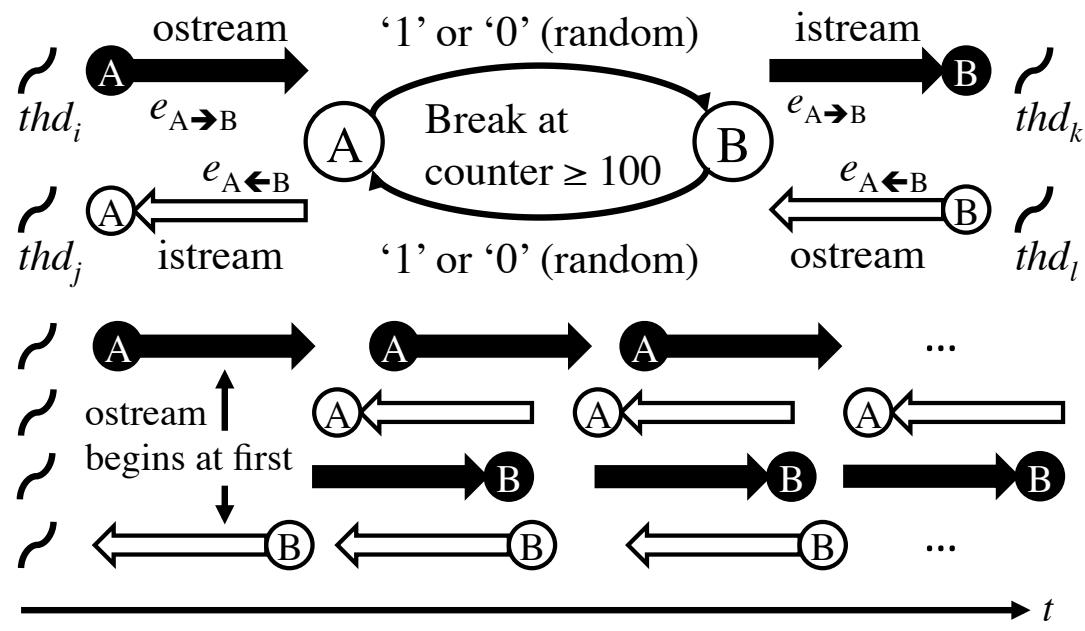
## □ Executor

- Submit your graph
- Debug your graph
- Execute your graph

# A concurrent ping-pong example



- A representative workload in parallel computing
  - Message passing back and forth concurrently
  - A fundamental building block of incremental flow



Method	Parallelism
C++17 thread	Local/ Distributed
MPI	Distributed

*Method to be compared with  
DtCraft*

# C++ thread on a local machine



## □ Standard C++ thread coding doesn't scale easily

```
auto count_A = 0;
auto count_B = 0;

// Boilerplate code to open file handles
int fds[2] = {-1, -1};
auto ret = socketpair(
    AF_UNIX,
    SOCK_NOBLOCK | SOCK_CLOEXEC,
    SOCK_STREAMfd,
    fds
);

if(ret == -1) {
    throw system_error("Failed on socketpair");
}

// Send a random binary data to fd and add the
// received data to the counter.
auto pingpong(int fd, int& count) {
    auto data = random<bool>()
    auto w = write(fd, &data, sizeof(data));
    if(w == -1 && errno != EAGAIN) {
        throw system_error("Failed on write");
    }
    data = 0;
    auto r = read(fds, &data, sizeof(data));
    if(r == -1 && errno != EAGAIN) {
        throw system_error("Failed on read");
    }
    count += data;
}

// Create a ping thread
thread t1 (
    [=] () {
        do {
            pingpong(fds[0], count_A);
        }while(count_A < 100);
    }
);

// Create a pong thread
std::thread t2 (
    [=] () {
        do {
            pingpong(fds[1], count_B);
        }while(count_B < 100);
    }
);

// Join the threads
t1.join();
t2.join();
```

*Amount of code grows with thread count and problem size!*

# C++ thread on distributed machines



## □ Things become massy going distributed ...

```
auto count_A = 0;
auto count_B = 0;

// Send a random binary data to fd and add the
// received data to the counter.
auto pingpong(int fd, int& count) {
    auto data = random<bool>()
    auto w = write(fd, &data, sizeof(data));
    if(w == -1 && errno != EAGAIN) {
        throw system_error("Failed on write");
    }
    data = 0;
    auto r = read(fds, &data, sizeof(data));
    if(r == -1 && errno != EAGAIN) {
        throw system_error("Failed on read");
    }
    count += data;
}

int fd = -1;
std::error_code errc;
```

*server.cpp*

```
if(getenv("MODE") == "SERVER") {
    fd = make_socket_server_fd("9999", errc);
}
else {
    fd = make_socket_client_fd("127.0.0.1", "9999", errc);
}

if(fd == -1) {
    throw system_error("Failed to make socket");
}
```

*client.cpp*

```
// Create a ping thread
thread t1 (
    []() {
        do {
            pingpong(fd, count_A);
        }while(count_A < 100);
    }
);

// Create a pong thread
std::thread t2 (
    []() {
        do {
            pingpong(fds[1], count_B);
        }while(count_B < 100);
    }
);

t1.join();
t2.join();
```

*Branch your code to server and client for distributed computation!*  
*simple.cpp → server.cpp + client.cpp (explicit and manual message passing)*

# Uh... you wonder how they look?



## □ make\_socket\_server\_fd and make\_socket\_client\_fd

```
int make_socket_server_fd(
    std::string_view port,
    std::error_code errc
) {
    int fd {-1};

    struct addrinfo hints;
    struct addrinfo* res {nullptr};
    struct addrinfo* ptr {nullptr};

    std::memset(&hints, 0, sizeof(struct addrinfo));
    hints.ai_family = AF_UNSPEC;
    hints.ai_socktype = SOCK_STREAM;
    hints.ai_protocol = IPPROTO_TCP;
    hints.ai_flags = AI_PASSIVE; // let it fill in port

    int one {1};
    int ret;

    if((ret = ::getaddrinfo(nullptr, port.data(), &hints)) != -1)
        errc = make_posix_error_code(ret);
    return -1;
}

// Try to connect to the first one that is available
for(ptr = res; ptr != nullptr; ptr = ptr->ai_next) {
    // Ignore undefined ip type.
    if(ptr->ai_family != AF_INET && ptr->ai_family != AF_INET6)
        goto try_next;
}

if((fd = ::socket(ptr->ai_family, ptr->ai_socktype,
    errc = make_posix_error_code(errno));
    goto try_next;
}

::setsockopt(fd, SOL_SOCKET, SO_REUSEADDR, &one, sizeof(one));

if(::bind(fd, ptr->ai_addr, ptr->ai_addrlen) == -1)
    errc = make_posix_error_code(errno);
    goto try_next;
}

if(::listen(fd, 128) == -1) {
    errc = make_posix_error_code(errno);
    goto try_next;
}
else {
    break;
}

try_next:
```

```
    if(fd != -1) {
        ::close(fd);
        fd = -1;
    }
}

::freeaddrinfo(res);

// Assign the socket to the underlying event native handle.
return fd;
```

```
make_fd_close_on_exec(fd);

tries = 3;

issue_connect:
ret = ::connect(fd, ptr->ai_addr, ptr->ai_addrlen);

if(ret == -1) {
    if(errno == EINTR) {
        goto issue_connect;
    }
    else if(errno == EAGAIN && tries--) {
        std::this_thread::sleep_for(std::chrono::milliseconds(500));
        goto issue_connect;
    }
    else if(errno != EINPROGRESS) {
        goto try_next;
    }
    errc = make_posix_error_code(errno);
}

// Poll the socket. Note that writable return doesn't mean it is connected.
if(select_on_write(fd, 5, errc) && !errc) {
    int optval = -1;
    socklen_t optlen = sizeof(optval);
    if(::getsockopt(fd, SOL_SOCKET, SO_ERROR, &optval, &optlen) < 0)
        errc = make_posix_error_code(errno);
    goto try_next;
}
if(optval != 0) {
    errc = make_posix_error_code(optval);
    goto try_next;
}
break;

try_next:
if(fd != -1) {
    ::close(fd);
    fd = -1;
}

::freeaddrinfo(res);

return fd;
```

*Actually more than the parallel code you need...*

# Massage Passing Interface (MPI)



## □ Explicitly move EVERYTHING between compute nodes

```
#define MAX_LEN 1 << 18 /* maximum vector length */
#define TRIALS 100 /* trials for each msg length */
#define PROC_0 0 /* processor 0 */
#define B0_TYPE 176 /* message "types" */
#define B1_TYPE 177

int numprocs, p, /* number of processors, proc index */
myid, /* this processor's "rank" */
length, /* vector length */
i, t;

double b0[MAX_LEN], b1[MAX_LEN]; /* vectors */

double start_time, end_time; /* "wallclock" times */

MPI_Status stat; /* MPI structure containing return */
/* codes for message passing operations */

MPI_Request send_handle, recv_handle; /* For nonblocking msgs */

MPI_Init(&argc,&argv); /* initialize MPI */

MPI_Comm_size(MPI_COMM_WORLD, &numprocs); /*how many processors? */
MPI_Comm_rank(MPI_COMM_WORLD, &myid); /*which one am I? */

if (myid == PROC_0)
{
    srand48(0xFEEDFACE);

    /* generate processor 0's vector */

    for (i=0; i<MAX_LEN; ++i)
        b0[i] = (double) drand48();
}

MPI_Barrier(MPI_COMM_WORLD);
```

*Hard-coded message passing*

```
for (length=1; length<=MAX_LEN; length*=2)
{
    MPI_Barrier(MPI_COMM_WORLD);

    if (myid == PROC_0)
        start_time = MPI_Wtime();

    for (t=0; t<TRIALS; ++t)
    {
        #ifdef BLOCKING
            if (myid == PROC_0)
            {
                MPI_Send(b0, length, MPI_DOUBLE, 1,
                        B1_TYPE, MPI_COMM_WORLD);
                MPI_Recv(b1, length, MPI_DOUBLE, 1,
                        B0_TYPE, MPI_COMM_WORLD, &stat);
            }
            else
            {
                MPI_Recv(b1, length, MPI_DOUBLE, 0,
                        B1_TYPE, MPI_COMM_WORLD, &stat);
                MPI_Send(b0, length, MPI_DOUBLE, 0,
                        B0_TYPE, MPI_COMM_WORLD);
            }
        #else
            MPI_Isend(b0, length, MPI_DOUBLE, (myid+1)%numprocs,
                      B0_TYPE, MPI_COMM_WORLD, &send_handle);
            MPI_Irecv(b1, length, MPI_DOUBLE, (myid+1)%numprocs,
                      B0_TYPE, MPI_COMM_WORLD, &recv_handle);

            MPI_Wait(&send_handle, &stat);
            MPI_Wait(&recv_handle, &stat);
        #endif
    }

    MPI_Finalize();
```

**DEADLOCK**

*It's user's fault to introduce deadlock*

# Concurrent ping-pong with DtCraft



```
auto Ball(Vertex& v, auto& k) {
    (*v.ostreams(k))((rand()%2));
    return Stream::DEFAULT;
};

auto PingPong(auto& v, auto& is, auto& k, auto& c) {
    int data;
    is(data);
    if((c+=data) >= 100) return Stream::REMOVE_THIS;
    else return Ball(v, k)
}

key_type AB, BA;
auto count_A {0}, count_B {0};

Graph G;
auto A = G.vertex().on([&](auto& v){ Ball(v, AB); });
auto B = G.vertex().on([&](auto& v){ Ball(v, BA); });

// Insert an iostream A->B
AB = G.stream(A, B).on(
    [&] (auto& B, auto& istream) {
        return PingPong(v, istream, BA, count_B);
});

// Insert an iostream B->A
BA = G.stream(B, A).on(
    [&] (auto& A, auto& istream) {
        return PingPong(v, istream, AB, count_A);
});

// Resource control using Linux container
G.containerize(A, "memory=1KB", "num_cpus=1");
G.containerize(B, "memory=1KB", "num_cpus=1");

Executor(G).dispatch();
```

- Fewer lines of code overall
- Less boilerplate code
- Single program
- No explicit data management
- Easy-to-use streaming interface
- Asynchronous by default
- Scalable to many threads
- Scalable to many machines
- In-context resource controls
- Scale out to heterogeneous devices
- Transparent concurrency controls
- Robust runtime via Linux container
- ... and more

# Be gentle to existing systems

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- ❑ **No one can claim their system general**
  - ❑ If yes, I understand it's for business purpose ☺
- ❑ **Big-data tools**
  - ✓ Good for data-driven and MapReduce workload
  - ✗ Bad for CPU/memory-intensive applications
- ❑ **High-performance computing (HPC) language**
  - ✓ Enabled the vast majority of HPC results for 20 years
  - ✗ Suffer from too many distinct notations for parallel programming
  - ✗ Analogous to assembly language (bottom-up design)
- ❑ **DtCraft**
  - ✓ A higher-level alternative to higher-level technologies
  - ✓ Transparent concurrency without taking away low-level controls
  - ✗ Currently best suitable for compute-intensive applications



# System implementation of DtCraft



## ❑ Kernel – *Master, Agent, and Executor*

- ❑ Master: global scheduling, deployment, and front-end
- ❑ Agent: local scheduling, containerization
- ❑ Executor: task execution (local, distributed, submitted modes)

## ❑ Event-driven programming environment

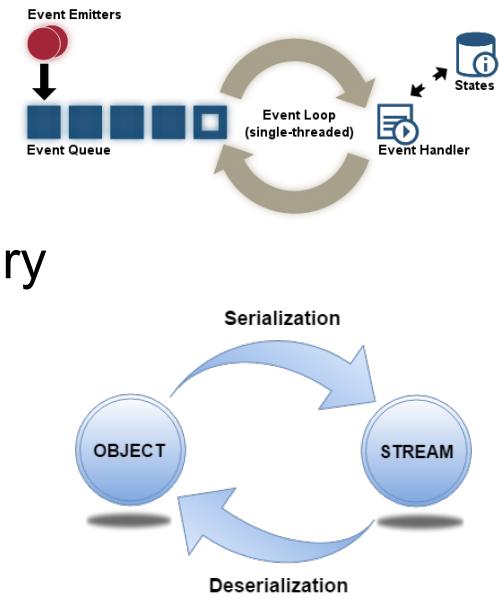
- ❑ Redesign the reactor library
- ❑ Thread-safe, lock-free, non-blocking IO

## ❑ Streaming interface

- ❑ Redesign the serialization/deserialization library
- ❑ Thread-safe, strongly typed, memory efficient

## ❑ Linux container

- ❑ A thin layer of fine-grained resource control
- ❑ Secure, safe, and robust



# A modern reactor library for event-driven programming



## □ The key component to our system kernel

```
// Smart pointer for effective concurrency control.
class Event : enable_shared_from_this<Event> {
    friend class Reactor;
    enum Type {TIMEOUT, PERIODIC, READ, WRITE};
    const function<Signal(Event&)> on;
};

class Reactor {

    // Executing event callback on a shared thread pool
    Threadpool threadpool;
    unordered_set<shared_ptr<Event>> eventset;

    template <typename T, typename... U>
    future<shared_ptr<T>> insert(U&&... u) {
        auto e = make_shared<T>(forward<U>(u)...);
        return promise(
            [&, e=move(e)](){
                _insert(e); // insert an event into reactor
                return e;
            }
        );
    }
};
```

- Written in C++17
- Thread-safe
- Lock-free
- Flattened event type
- Task-based parallelism
- Single producer (promise)
- Multiple consumers (future)
- Smart pointer
- Non-blocking IO controls
- Support multiple back-ends
- Shared thread pool
- Callback in a critical section

# A memory-efficient serialization/deserialization library



## □ The key component to our message passing

```
struct MyData {
    int raw;
    std::string str;
    std::vector<int> vec;

    // All you need is to include necessary members
    template <typename Archiver>
    auto archive(Archiver& ar) {
        return ar(raw, str, vec); // we use variadic template
    }
};

class BinaryOutputArchiver {

    OutputStreamBuffer& osbuf;

    template <typename... U>
    constexpr streamsize operator()((U&&... u) {

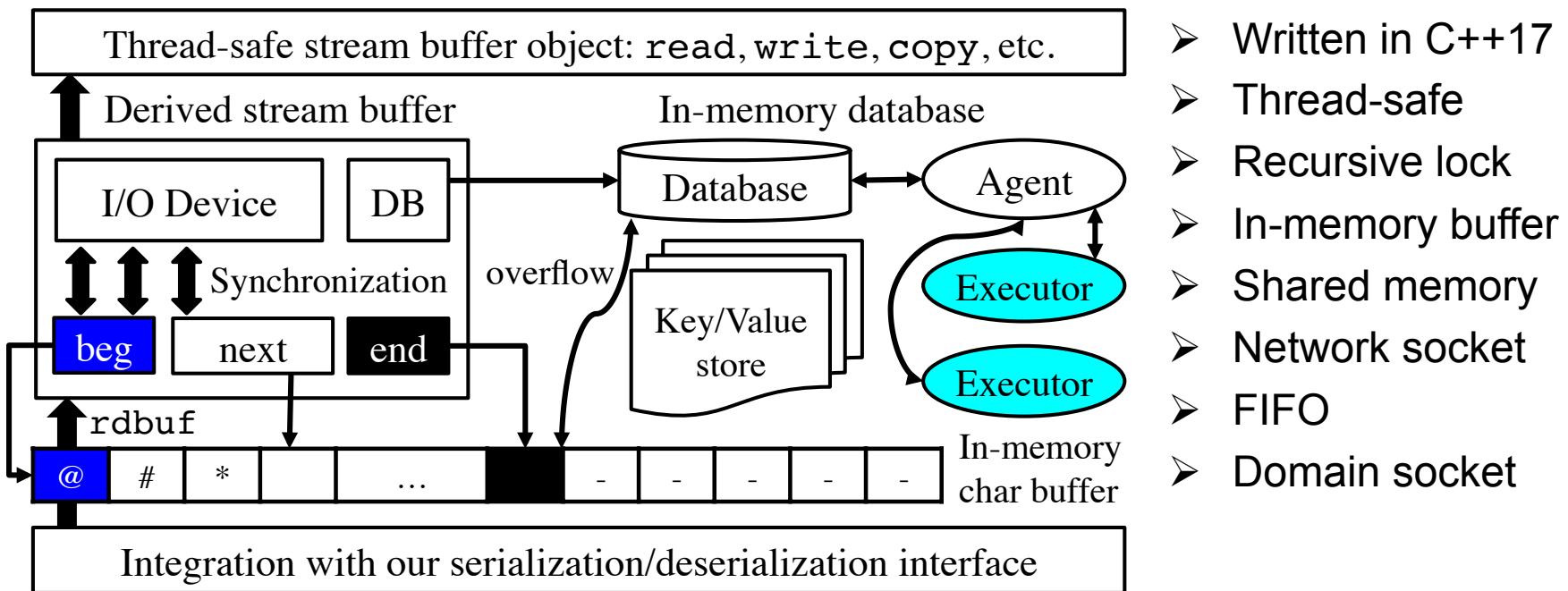
        if constexpr (is_POD<U>) {
            osbuf.write(&u, sizeof(u));
        }
        else if constexpr (is_std_string<U>) {
            osbuf.write(u.data(), u.size());
        }
        else if constexpr (is_std_vector<U>) {
            // ...
        }
        // ...
        else
            return archive(forward<U>(u)...);
    }
};
```

- Written in C++17
- Heavy meta-programming
- Thread-safe
- Strongly-typed
- Convenient to use
- Integrated with our IO buffer
- Binary data format
- No extra parsing/unpacking
- No secondary representation
- Memory-efficient
- STL ready-to-use

# Concurrent input/output stream buffer



- In charge of reading/writing operations on devices
  - Work directly with our serialization/deserialization interface
  - Zero copy in user space



# A Linux container-based resource control



## ☐ Namespace isolation & resource control

```
// Linux container
class ContainerDescriptor {

    friend class Graph;

    const key_type key;
    operator key_type() const;

    ContainerDescriptor& add(key_type);
    ContainerDescriptor& num_cpus(unsigned);
    ContainerDescriptor& memory_limit_in_bytes(uintmax_t);
    ContainerDescriptor& blkio(uintmax_t);
    // ...
};

auto A = G.vertex();
auto B = G.vertex();

// Create a container for vertex A with 1 CPU,
// 1 MB memory, and 1 GB block IO
G.container().add(A)
    .num_cpus(1)
    .memory_limit_in_bytes(1_MB)
    .blkio(1_GB);

// Create a container for vertex B with 2 CPU,
// 2 MB memory, and 2 GB block IO
G.container().add(B)
    .num_cpus(2)
    .memory_limit_in_bytes(2_MB)
    .blkio(2_GB);
```

- Safe and robust runtime
- Minimize intruder's effect
- Network isolation
- UTS isolation
- IPC isolation
- PID isolation
- User/Group isolation
- Cgroup isolation
- Mount point isolation
- In-context resource controls
- Give scheduler hints
- Maximize cluster performance

# Graph deployment and workload distribution

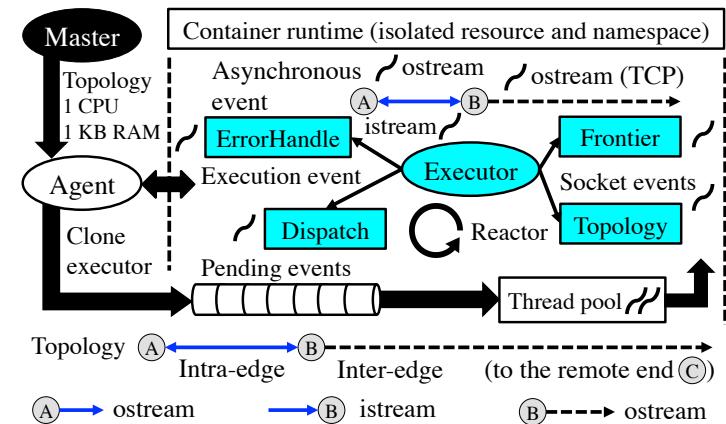
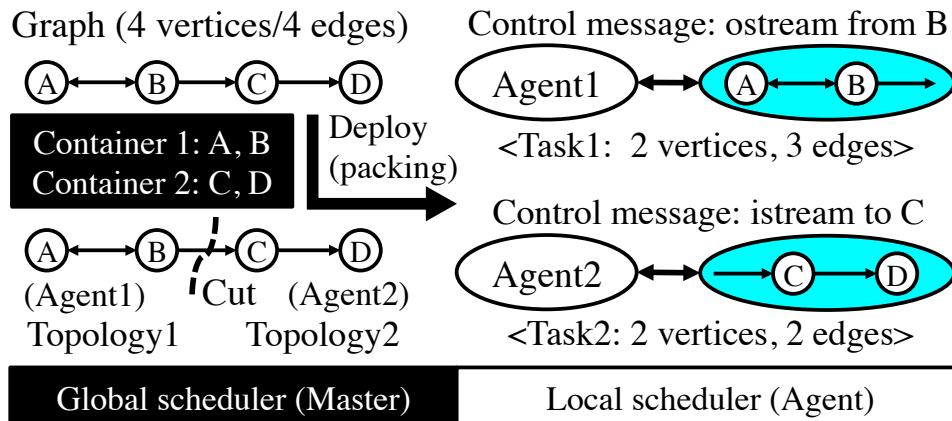


## □ Global scheduler – master

- Manage users' graph submissions
- Partition graph through bin-packing optimization

## □ Local schedulers – agents

- Fork-exec an executor for each topology
- Containerize the executor under resource constraints

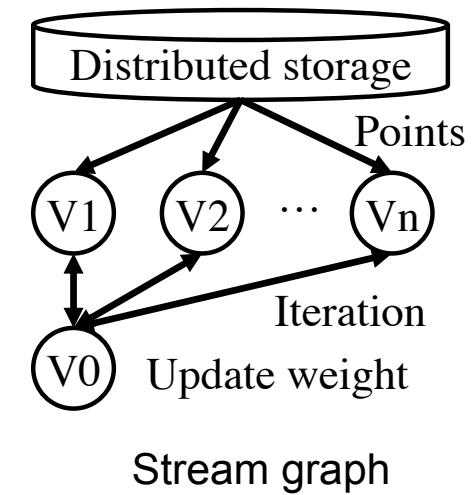
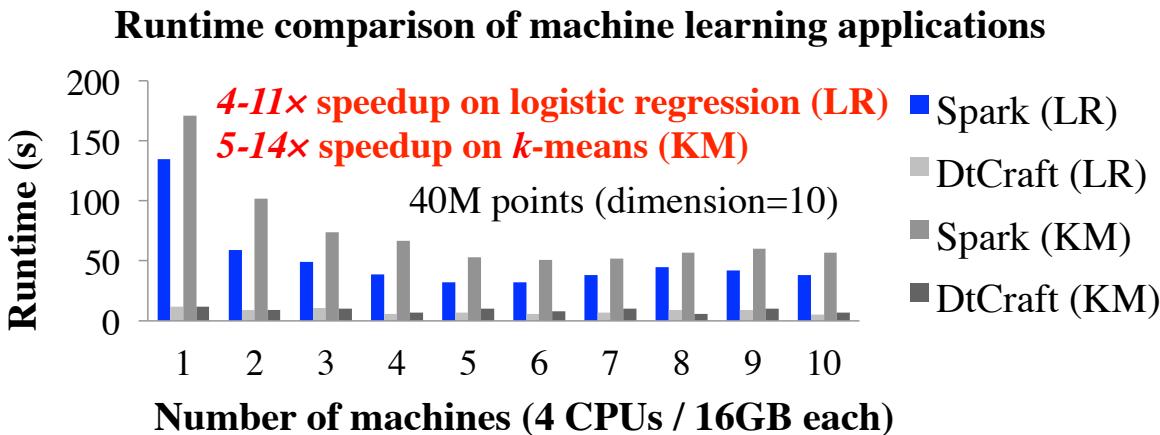


*Intra-stream and inter-stream talk through shared memory and TCP socket, respectively*

# Experiments on machine learning



- Logistic regression and *k*-means algorithms
  - Mimic the MapReduce-based flow with ten iterations
- Compared with Spark 2.0 MLib
  - More than an order of magnitude faster
  - No extra overhead on the first iteration to cache data
  - Explicit resource controls outperform blind RDD partitions



# Experiments on graph algorithms

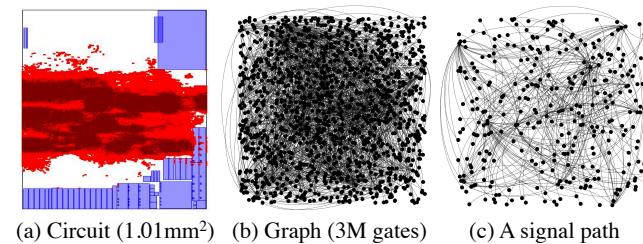


## □ Shortest path algorithm

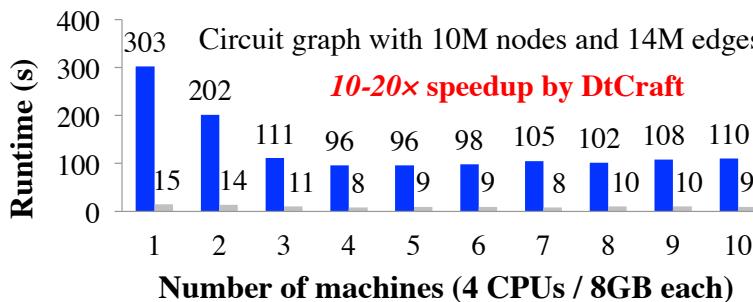
- Circuit graph with 10M nodes and 14M edges
- Higher connectivity than many of big data graphs
- Mimic the Pregel-based algorithm (Bellman-Ford style)

## □ Compared with Spark 2.0 GraphX

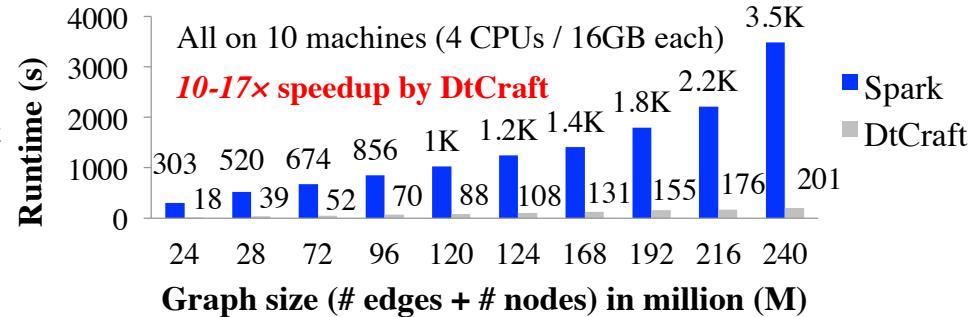
- Less synchronization overhead
- An order of magnitude faster
- Scale up as the graph size increases



Runtime comparison of shortest path finding



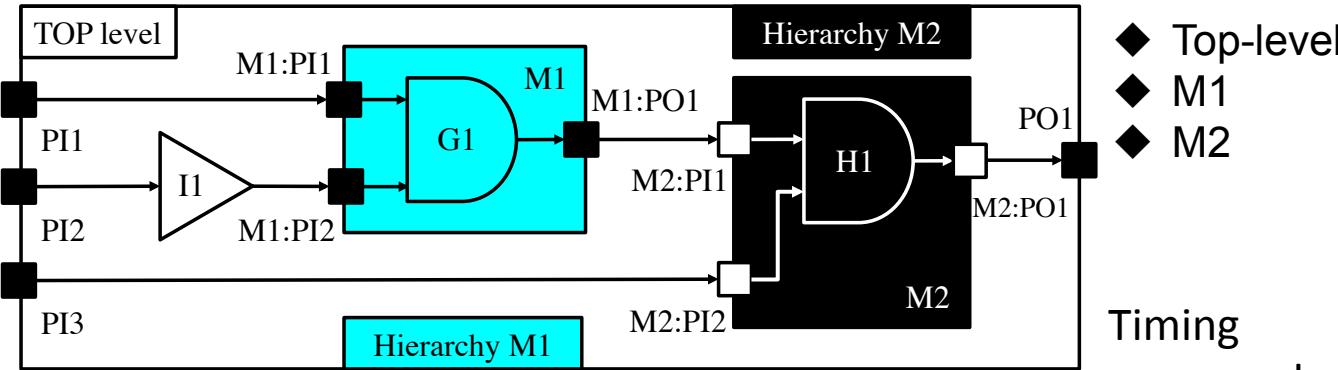
Performance scalability (runtime vs graph size)



# Distributed timing analysis using DtCraft

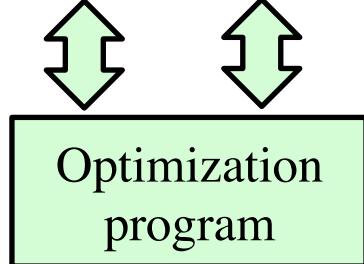
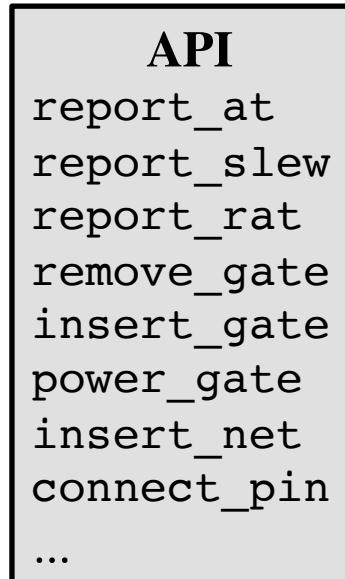
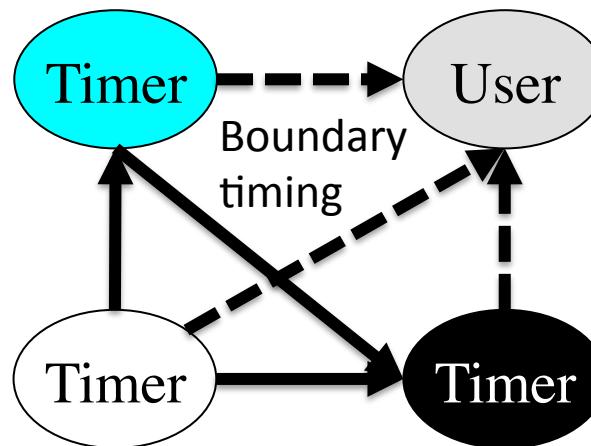


## □ Two-level hierarchical design (three partitions)



- Three timer vertices
- One user vertex
- Four Linux containers
- Six input/output streams

*Each container has one OpenTimer operating on one design hierarchy*



# Exchange timing data – delay, slew, etc.



## DtCraft

```
// Timing data (early/late rise/fall)
struct Timing {
    string pin;
    array<float, 4> value;

    template <typename T>
    auto archive(T& ar) {
        return ar(pin, value);
    }
};

// Timing path
struct Path {
    float slack;
    vector<string> pins;

    template <typename T>
    auto archive(T& ar) {
        return ar(slack, pins);
    }
};

// Exchange timing through DtCraft stream
stream.on(
    [] (Vertex& v, InputStream& is) {
        if(Timing timing; is(timing) != -1) {
            // Call OpenTimer to run incremental timing
        }
    }
);
```

*In-context streaming with < 30 lines*

## Existing framework

Hard-code your message format

```
// OpenTimer.proto
package OpenTimer;

// Message format for timing
message Timing {
    required string pin = 0;
    required float er = 1;
    required float ef = 2;
    required float lr = 3;
    required float lf = 4;
}

// Message format for path
message Path {
    required float slack = 0;
    repeated string pins = 1;
}
```

*Many extra stuff ☹*

**Extra.pb.h**

**Extra.pb.cpp**

...

**Source.cpp**

Google Protocol Buffer  
(open-source compiler)

C++/Java/Python  
source code generator

**.cpp/.h class methods**

ParseFromArray(void\*, size\_t)  
SerializeToArray(void\*, size\_t)

Message wrapper

**Derived packet struct**

header\_t header  
void\* buffer



*Out-of-context streaming takes > 300 lines*

# Deploy the distributed timer in one line



DtCraft



Existing framework

```
// Create a timer vertex for Top
auto Top = G.Vertex().on(
    [=] () {
        OpenTimer timer ("Top.v");
    }
);

// Create a timer vertex for Macro 1
auto M1 = G.Vertex().on(
    [=] () {
        OpenTimer timer ("M1.v");
    }
);

// Create a timer vertex for Macro 2
auto M2 = G.Vertex().on(
    [=] () {
        OpenTimer timer ("M2.v");
    }
);

// Create streams ...
...

// Distribute timers to machines.
G.container().add(Top).num_cpus(4).memory_(4_GB);
G.container().add(M1).num_cpus(1).memory(8_GB);
G.container().add(M2).num_cpus(2).memory(6_GB);

~$ ./submit --master=127.0.0.1 binary
```

*Only three lines for  
resource control in  
Linux container*

*Duplicate the code for each partition*

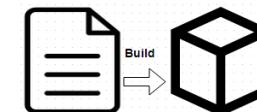
Top.cpp



M1.cpp



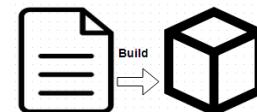
M2.cpp



Dockerfile



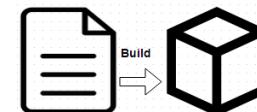
Image



Dockerfile



Image



Dockerfile



Image

Container 1

Container 2

Container 3

*Wrap up with submission scripts*



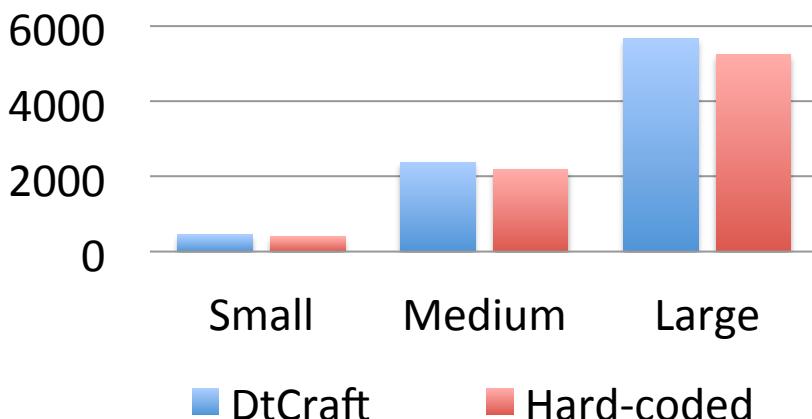
# Comparison with the hard-coded method



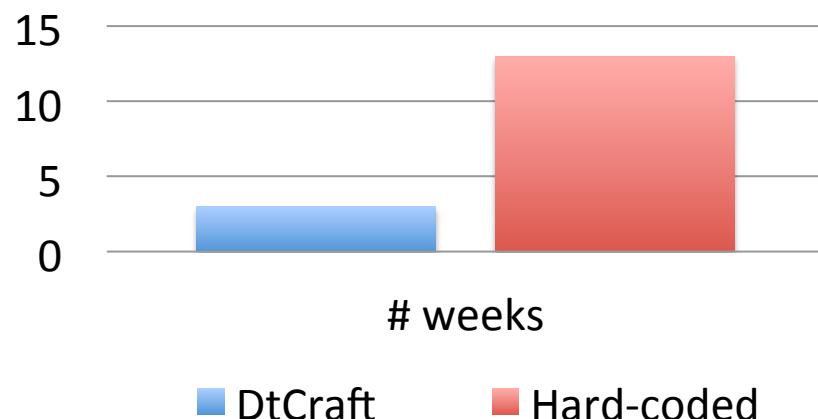
- **×17 fewer lines of code**
  - 33% from message passing
  - 67% from boilerplate code
- **7-11% performance loss**
  - Transparent concurrency
  - API cost

*"With DtCraft, it took me only three weeks, precisely, the **SPARE time** out of my summer internship at Citadel, to build a distributed timer that otherwise took my **whole summer internship** with IBM".*

**Runtime (40 AWS nodes)**



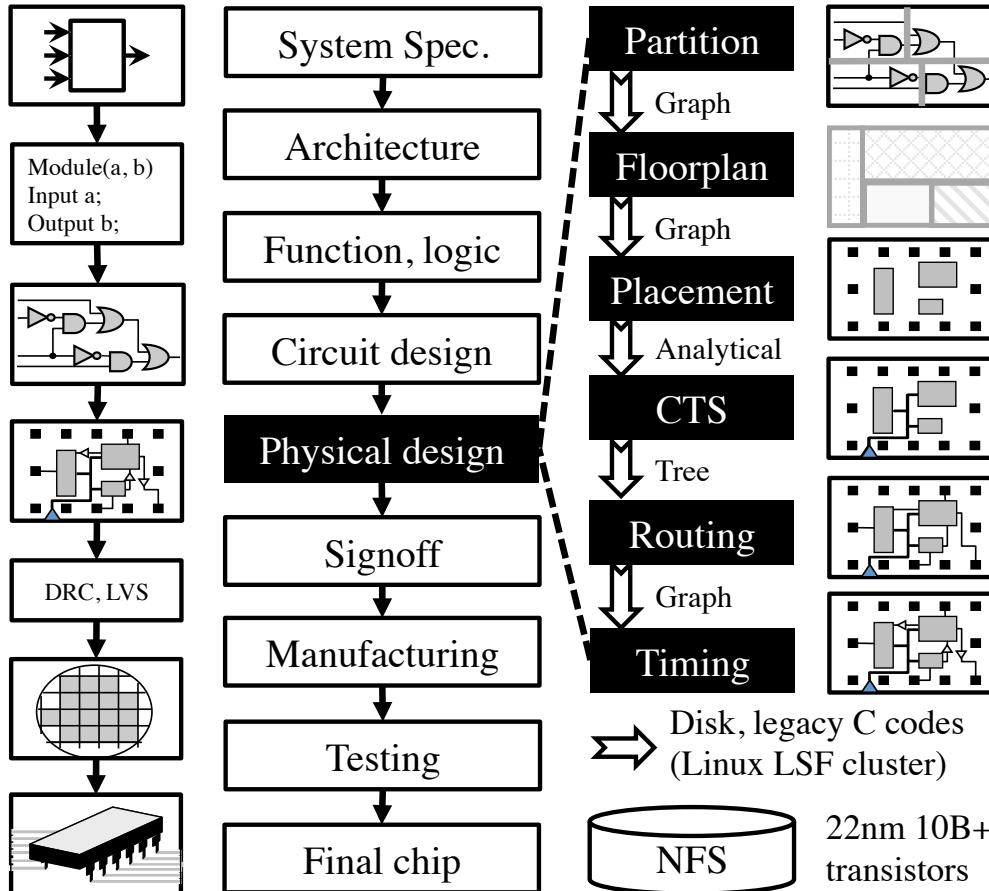
**Development time**



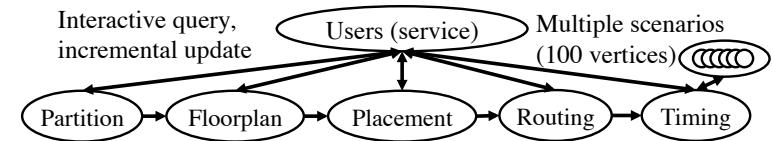
# Experiments on EDA tool Integration



## □ Electronic design automation (EDA)



Stream graph for physical design



## □ Goal

- New EDA methodology
- distributed integration
- Reduce tool-to-tool overhead

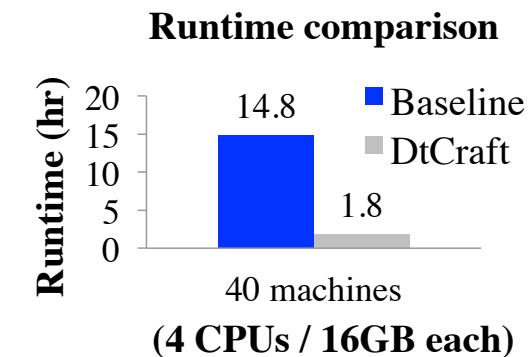
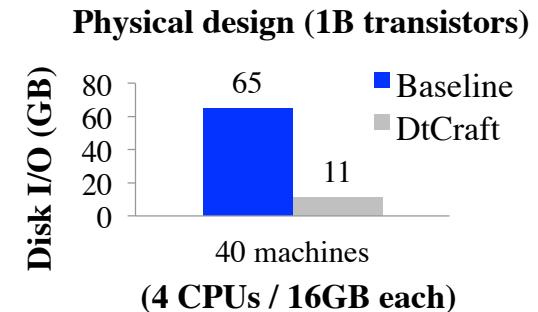
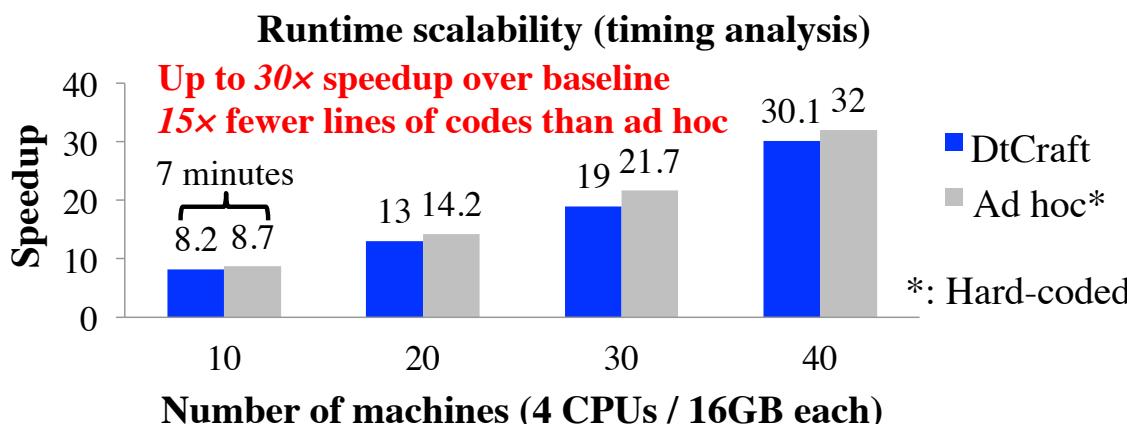
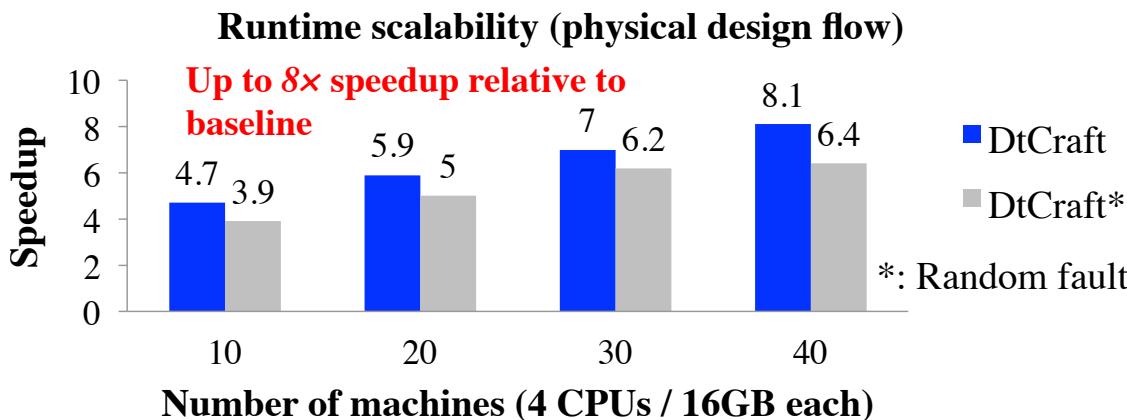
## □ Open-source tools

- OpenTimer, placer, etc.

# Experiments on EDA tool integration (cont'd)



## □ Physical design and timing analysis



*Less disk IO translates to faster runtime*

# Conclusion

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- ❑ **DtCraft: A distributed execution engine**
  - ❑ Creation of new parallel/distributed algorithms
  - ❑ Tool-to-tool integration at cloud scale
- ❑ **Tentative first release on 12/1**
  - ❑ Github repository
- ❑ **Acknowledgment**
  - ❑ UIUC CAD group

# Thank you!

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*Boost your productivity in writing parallel code!*