

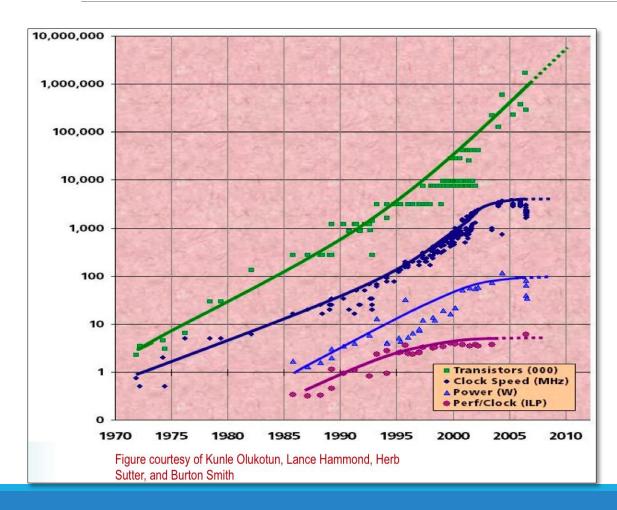
Managing Asynchrony in C++

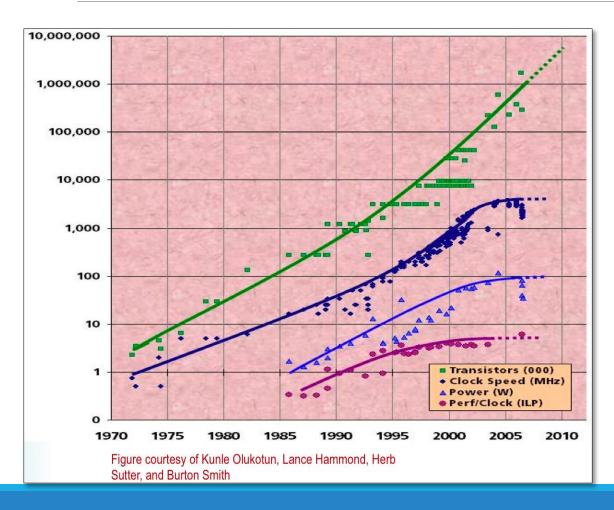
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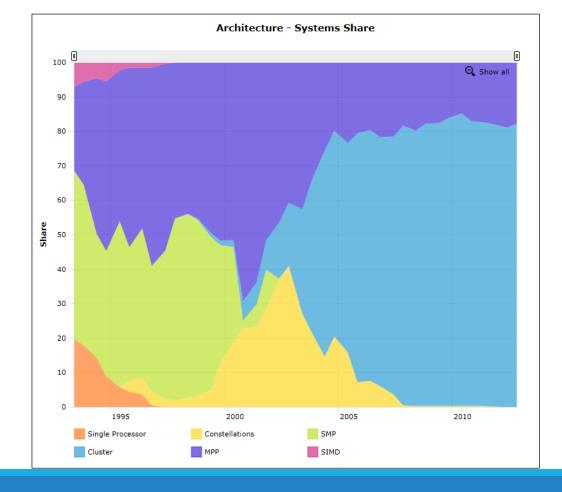
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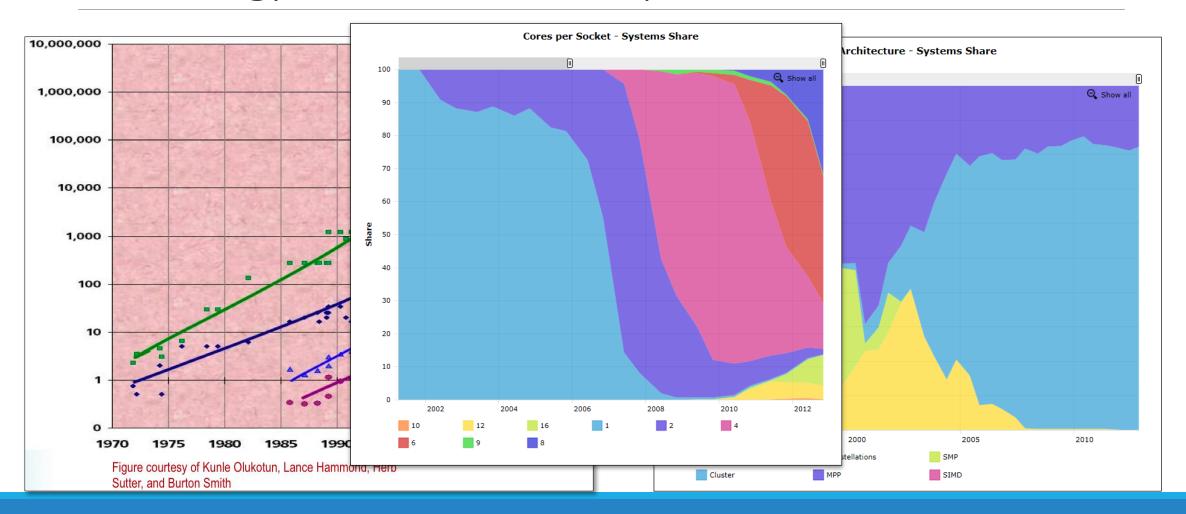
The Venture Point

TECHNOLOGY DEMANDS NEW RESPONSE











Peak performance: 1.17 PetaFLOPs

112,896 computing cores (18,816 2.6 GHz six-core AMD Opteron processors)

Amdahl's Law (Strong Scaling)

$$S = \frac{1}{(1-P) + \frac{P}{N}}$$

S: Speedup

P: Proportion of parallel code

N: Number of processors

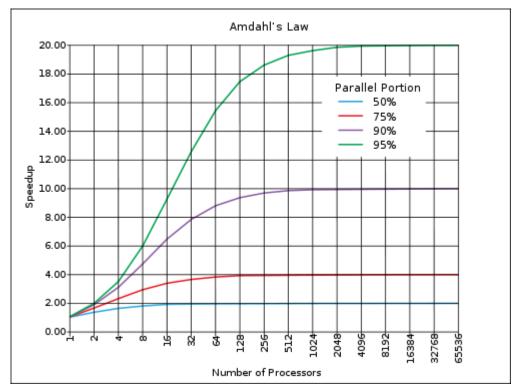


Figure courtesy of Wikipedia (http://en.wikipedia.org/wiki/Amdahl's_law)

The 4 Horsemen of the Apocalypse: **SLOW**

Starvation

 Insufficient concurrent work to maintain high utilization of resources

Latencies

Time-distance delay of remote resource access and services

Overheads

 Work for management of parallel actions and resources on critical path which are not necessary in sequential variant

Waiting for Contention resolution

 Delays due to lack of availability of oversubscribed shared resources



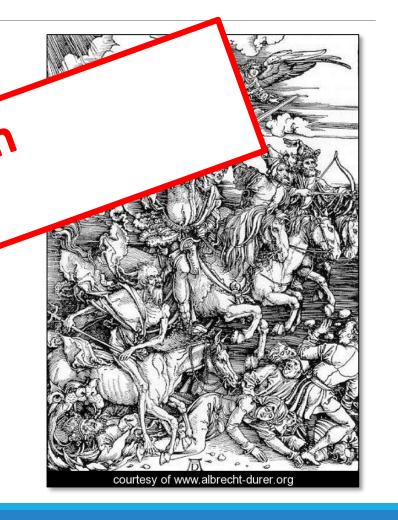
The 4 Horsemen of the Apocalypse: **SLOW**

Starvation

 Insufficient concurrent work to maintain high utilization of resources

Overhead upper bound on both scaling scaling waveak and strong scaling waveak and strong scaling or necessary

De vs due to lack of availability of oversubscribed shared resources



The Challenges

We need to find a usable way to <u>fully</u> parallelize the applications

Goals are

- Defeat The Four Horsemen
- Provide manageable paradigms for handling parallelism
- Expose asynchrony to the programmer without exposing concurrency
- Make data dependencies explicit, hide notion of 'thread', 'communication', and 'data distribution'

Runtime Systems

THE NEW DIMENSION

HPX – A General Purpose Runtime System

All examples in this talk are based on HPX

A general purpose runtime system for applications of any scale

- http://stellar.cct.lsu.edu/
- https://github.com/STEIIAR-GROUP/hpx/

Exposes an uniform, standards-oriented API for ease of programming parallel and distributed applications.

Enables to write fully asynchronous code using hundreds of millions of threads.

Provides unified syntax and semantics for local and remote operations.

Is published under Boost license and has an open, active, and thriving developer community.

HPX – A General Purpose Runtime System

Governing principles

- Active global address space (AGAS) instead of PGAS
- Message driven instead of message passing
- Lightweight control objects instead of global barriers
- Latency hiding instead of latency avoidance
- Adaptive locality control instead of static data distribution
- Moving work to data instead of moving data to work
- Fine grained parallelism of lightweight threads instead of Communicating Sequential Processes (CSP/MPI)

Fully asynchronous

- All possibly remote operations are asynchronous by default
 - 'Fire & forget' semantics (result is not available)
 - 'Pure' asynchronous semantics (result is available via hpx::future)
- Composition of asynchronous operations (N3634)
 - hpx::when_all, hpx::when_any, hpx::when_n
 - hpx::future::then(f)
- Can be used 'synchronously', but does not block
 - Thread is suspended while waiting for result
 - Other useful work is performed transparently

As close as possible to C++11 standard library, where appropriate, for instance

- ∘ std::thread → hpx::thread
- ∘ std::mutex → hpx::mutex
- ∘ std::future → hpx::future
- ∘ std::async → hpx::async
- ∘ std::bind → hpx::bind
- ∘ std::function → hpx::function
- ∘ std::tuple → hpx::tuple
- std::any \rightarrow hpx::any (N3508)
- ∘ std::cout → hpx::cout
- etc.

Fully move enabled (using Boost.Move)

hpx::bind, hpx::function, hpx::tuple, hpx::any

Fully type safe remote operation

- Extends the notion of a 'callable' to remote case (actions)
- Everything you can do with functions is possible with actions as well

Data types are usable in remote contexts

- Can be sent over the wire (hpx::bind, hpx::function, hpx::any)
- Can be used with actions (hpx::async, hpx::bind, hpx::function)

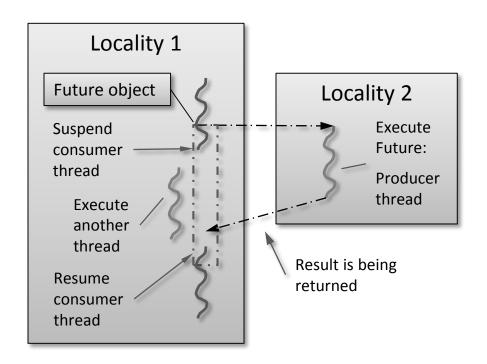
R f(p)	Synchronous (return R)	Asynchronous (return future <r>)</r>	Fire & Forget (return void)
Functions (direct)	f(p) C++	async(f, p)	apply(f, p)
Functions (lazy)	bind(f, p)()	<pre>async(bind(f, p),)</pre>	apply(bind(f, p),)
Actions (direct)	HPX_ACTION(f, a) a(id, p)	<pre>HPX_ACTION(f, a) async(a, id, p)</pre>	<pre>HPX_ACTION(f, a) apply(f, id, p)</pre>
Actions (lazy)	<pre>HPX_ACTION(f, a) bind(a, id, p)()</pre>	<pre>HPX_ACTION(f, a) async(bind(a, id, p),)</pre>	<pre>HPX_ACTION(f, a) apply(bind(a, id, p),) HPX</pre>

The Future

A CLOSER LOOK

What is a (the) future

A (std) future is an object representing a result which has not been calculated yet



- Enables transparent synchronization with producer
- Hides notion of dealing with threads
- Makes asynchrony manageable
- Allows for composition of several asynchronous operations
- Turns concurrency into parallelism

What is a (the) Future?

Many ways to get hold of a future, simplest way is to use (std) async:

```
int universal answer() { return 42; }
void deep_thought()
    future<int> promised_answer = async(&universal_answer);
    // do other things for 7.5 million years
    cout << promised answer.get() << endl; // prints 42</pre>
```

Stupidest Way to Calculate Fibonacci Numbers

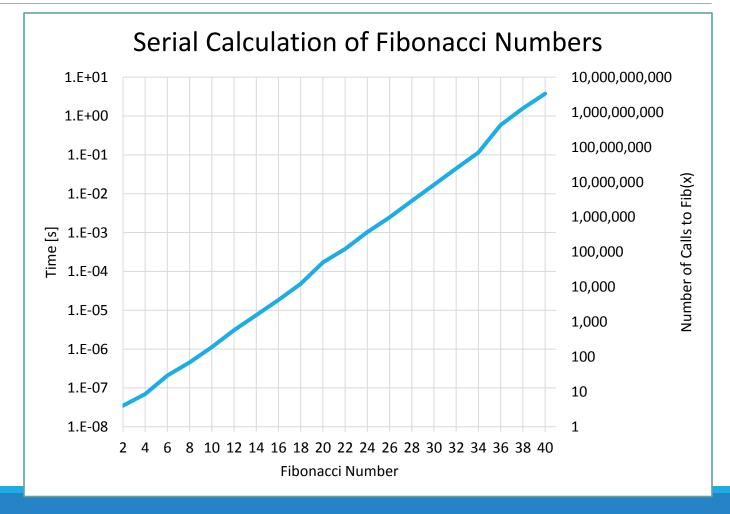
Synchronous way:

```
// watch out: O(2^n)
int fibonacci_serial(int n)
    if (n < 2) return n;
    return fibonacci_serial(n-1) + fibonacci_serial(n-2);
cout << fibonacci_serial(10) << endl; // will print: 55</pre>
```

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Stupidest Way to Calculate Fibonacci Numbers

Complexity: O(2ⁿ)



Stupidest Way to Calculate Fibonacci Numbers

Computational complexity is $O(2^n)$ – alright, however

This algorithm is representative for a whole class of applications

- Tree based recursive data structures
 - Adaptive Mesh Refinement important method for wide range of physics simulations
 - Game theory

5/14/2013

- Graph based algorithms
 - Breadth First Search

Characterized by very tightly coupled data dependencies between calculations

But fork/join semantics make it simple to reason about parallelization

Let's spawn a new thread for every other sub tree on each recursion level

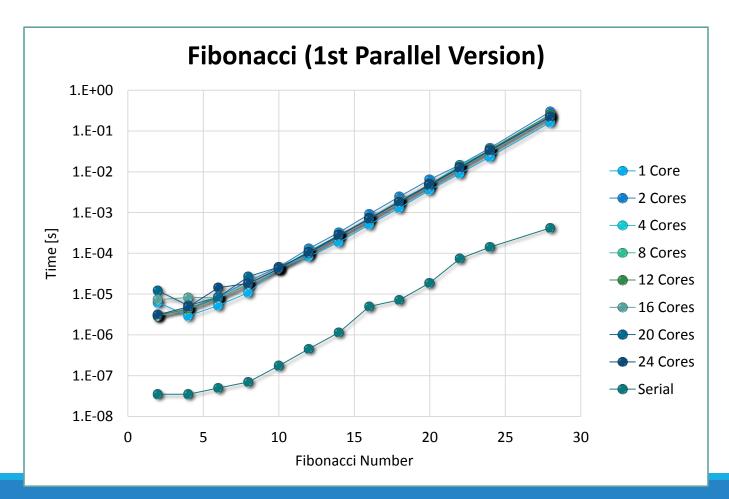
Explicit Asynchrony

Let's Parallelize It! What could be easier?

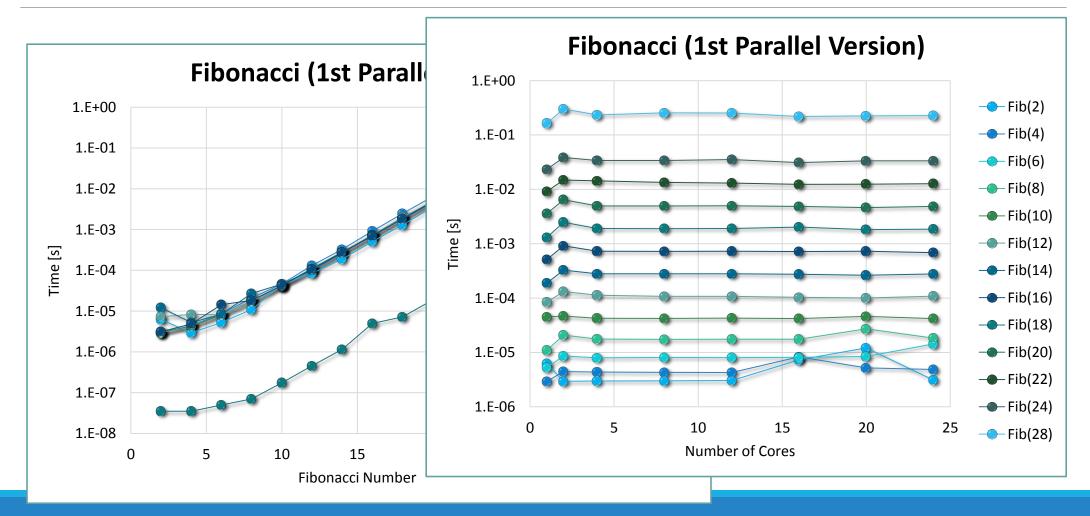
Calculate Fibonacci numbers in parallel (1st attempt)

```
uint64_t fibonacci(uint64_t n)
   // if we know the answer, we return the value
   if (n < 2) return n;
   // asynchronously delay-calculate one of the sub-terms
   future<uint64 t> f = async(launch::deferred, &fibonacci, n-1);
   // synchronously calculate the other sub-term
   uint64_t r = fibonacci(n-2);
   // wait for the future and calculate the result
    return f.get() + r;
```

Let's Parallelize It! What could be easier?



Let's Parallelize It! What could be easier?



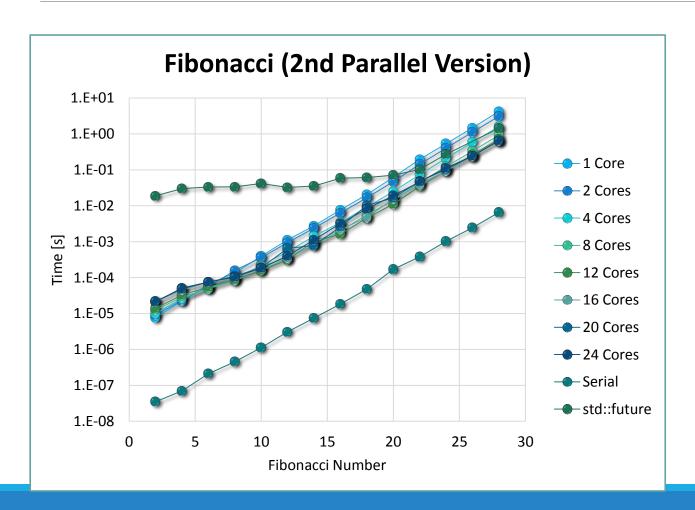
Let's Parallelize It! Could it be easier?

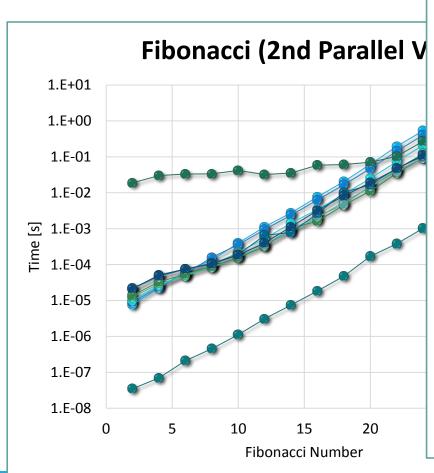
Parallel calculation (1st attempt), why is it slow? Why doesn't it scale?

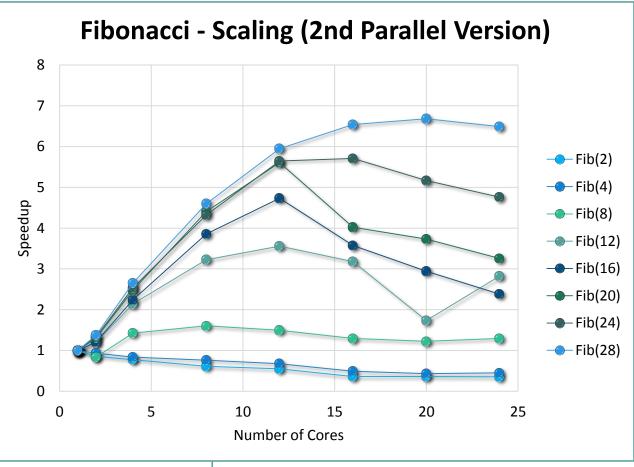
```
uint64 t fibonacci(uint64 t n)
   // if we know the answer, we return the value
   if (n < 2) return n;
   // asynchronously delay-ca<del>lculate one of the</del> sub-terms
   future<uint64_t> f = async(launch::deferred, &fibonacci, n-1);
   // synchronously calculate the other sub-term
    uint64 t r = fibonacci(n-2);
                  he future and calculate the result
    return f.get() + r;
```

Calculate Fibonacci numbers in parallel (2nd attempt)

```
uint64_t fibonacci(uint64_t n)
   // if we know the answer, we return the value
   if (n < 2) return n;
   // asynchronously calculate one of the sub-terms
   future<uint64 t> f = async(launch::async, &fibonacci, n-1);
   // synchronously calculate the other sub-term
   uint64_t r = fibonacci(n-2);
   // wait for the future and calculate the result
    return f.get() + r;
```







What's wrong? While it scales it is still 100 times slower than the serial execution

Creates a new future for each invocation of fibonacci() (spawns an HPX thread)

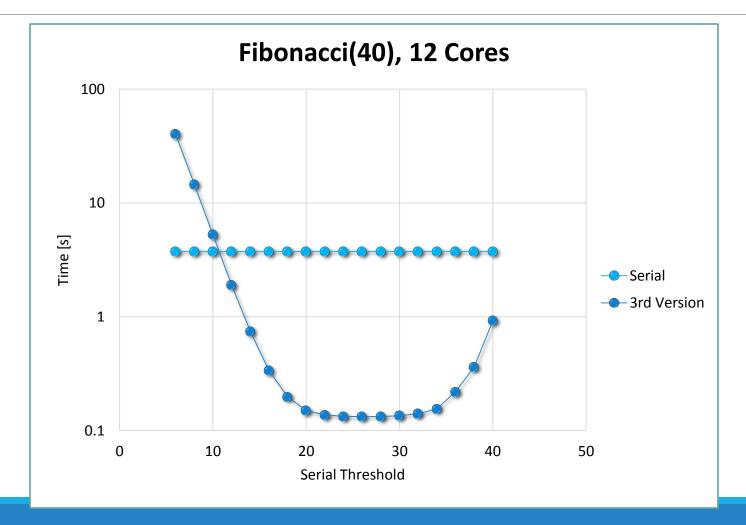
- Millions of threads with minimal work each
- Overheads of thread management (creation, scheduling, execution, deletion) are much larger than the amount of useful work
 - Future overheads: ~1μs (Thread overheads: ~400ns)
 - Useful work: ~50ns

Let's introduce the notion of granularity of work (grain size of work)

The amount of work executed in one thread

Parallel calculation (3rd attempt), switching to serial execution below given threshold

```
uint64_t fibonacci(uint64_t n)
    if (n < 2) return n;
    if (n < threshold) return fibonacci serial(n);</pre>
   // asynchronously calculate one of the sub-terms
   future<uint64 t> f = async(launch::async, &fibonacci, n-1);
    // synchronously calculate the other sub-term
    uint64_t r = fibonacci(n-2);
    // wait for the future and calculate the result
    return f.get() + r;
```



Granularity Control The New Dimension

Parallelizing code introduces Overheads (SLOW)

Overheads are caused by code which

- Is executed in the parallel version only
- Is on the critical path (we can't 'hide' it behind useful work)
- Is required for managing the parallel execution
 - i.e. task queues, synchronization, data exchange
 - NUMA and core affinities

Controlling not only the amount of resources used but also the granularity of work is an important factor

Controlling the grain size of work allows finding the sweet-spot between too much overheads and too little parallelism

N3634: Improvements to std::future<T> and Related APIs

Combining futures

- when_all(), when_any()
 - Allows waiting for a combination of passed in future instances
 - Return a future representing the entire operation

Defining continuations

- future::then()
 - Attaches a function to be executed once the future gets ready
 - Returns a future representing the result of the continuation function

Unwrapping futures

- Asynchronous operations may return future<future<T> >
 - future::unwrap() returns inner future instance

Create futures which are 'ready'

future<decay<T>::type> make_ready_future(T);

http://www.open-std.org/jtc1/sc22/wg21/docs/papers/2013/n3634.pdf

Futurization

Special technique allowing to automatically transform code

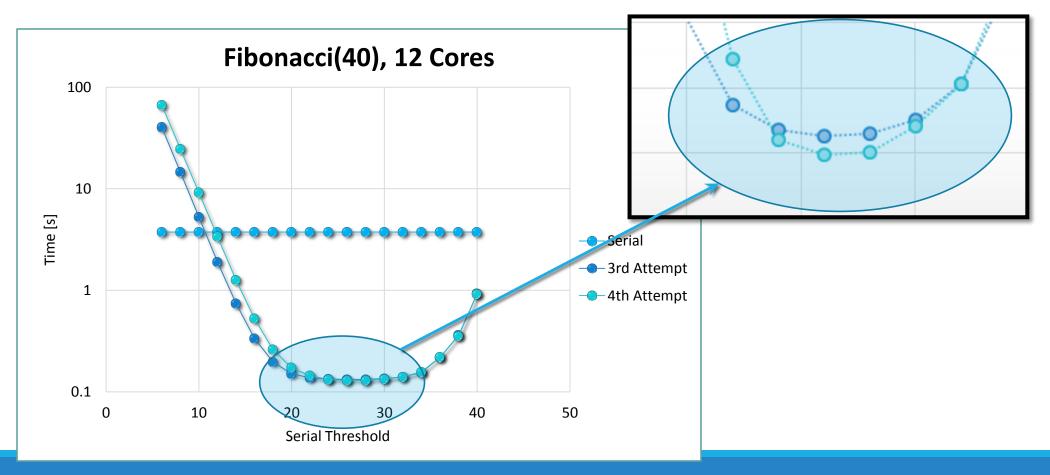
- Delay direct execution in order to avoid synchronization
- Turns 'straight' code into 'futurized' code
- Code no longer calculates results, but generates an execution tree representing the original algorithm
- If the tree is executed it produces the same result as the original code
- The execution of the tree is performed with maximum speed, depending only on the data dependencies of the original code
- Simple transformation rules:

Straight Code	Futurized Code
T func() {}	<pre>future<t> func() {}</t></pre>
rvalue: n	<pre>make_ready_future(n)</pre>
T n = func();	<pre>future<t> n = async(&func,);</t></pre>

```
future<uint64 t> fibonacci(uint64 t n)
    if (n < 2) return make_ready_future(n);</pre>
    if (n < threshold) return make ready future(fibonacci serial(n));
    future<future<uint64 t>> f = async(launch::async, &fibonacci, n-1);
    future<uint64 t> r = fibonacci(n-2);
    return when_all(f.get() , r).then(
        [](future<std::vector<future<uint64 t>>> fv) -> uint64 t {
            std::vector<future<uint64_t>> v = fv.get();
            return v[0].get() + v[1].get();
        });
```

```
future<std::vector<future<uint64_t>>> f = when_all(f.get(), r);
future<uint64 t> result = f.then(
    [](future<std::vector<future<uint64 t>>> fv) -> uint64 t {
        std::vector<future<uint64_t>> v = fv.get();
        return v[0].get() + v[1].get();
    });
return result;
```

```
future<uint64_t> fibonacci(uint64_t n)
    if (n < 2) return make_ready_future(n);</pre>
    if (n < threshold) return make_ready_future(fibonacci_serial(n));</pre>
    future<future<uint64 t>> f = async(launch::async, &fibonacci, n-1);
    future<uint64 t> r = fibonacci(n-2);
    return when_all(f.get(), r).then(
        [](future<std::vector<future<uint64_t>>> f) -> uint64_t {
            std::vector<future<uint64_t>> v = f.get();
            return v[0].get() + v[1].get();
        });
```



```
future<uint64_t> fibonacci(uint64 t n)
    if (n < 2) return make_ready_future(n);</pre>
    if (n < threshold) return make_ready_future(fibonacci_serial(n));</pre>
    future<future<uint64 t>> f = async(launch::async, &fibonacci, n-1);
    future<uint64 t> r = fibonacci(n-2);
    return when_all(f.get(), r).then(
        [](future<std..vector<future<uint64_t>>> f) -> uint64_t {
            std::vector<future<uint64_t>> v = f.get();
            return v[0].get() + v[1].get();
        });
```

Parallel way (5th attempt), unwrapping inner future

```
future<uint64_t> fibonacci(uint64_t n)
    if (n < 2) return make_ready_future(n);</pre>
    if (n < threshold) return make_ready_future(fibonacci_serial(n));</pre>
    future<uint64 t> f = async(launch::async, &fibonacci, n-1).unwrap();
    future<uint64 t> r = fibonacci(n-2);
    return when all(f, r).then(
        [](future<std::vector<future<uint64_t> > > f) -> uint64_t {
            std::vector<future<uint64_t> > v = f.get();
            return v[0].get() + v[1].get();
        });
```

Guess what? – This is the fastest implementation so far!

Large amount of overheads in HPX are caused by stacks

- We don't know which thread will suspend, thus every thread needs its own stack segment (~8kByte)
 - Virtual memory segmentation, TLB thrashing, physical memory exhaustion
- Not always possible to reuse stack segments as too many threads are being suspended
 - Fibonacci: many threads get created just to be suspended almost immediately
 - Figuring out what threads need to be executed first in order to make progress is a NP complete problem

Using when_all(), then(), etc. is an alternative, however

- Complex constructs
- Integrate badly with straight serial code and control structures

HPX can 'simulate' synchrony while performing asynchronous operations:

```
int f(stream str)
{
    std::vector<char> buf;
    future<int> count = str.read(512, buf);
    // ...
    return count.get() + 11;  // get() will suspend
}
```

However, this leaves a half-filled stack frame behind, moreover caller can't proceed until done

Introduces 2 new keywords: async and await

```
future<int> f(stream str) async
{
    shared_ptr<vector<char>> buf = ...;
    int count = await str.read(512, buf);  // returns from f() if not ready!
    return count + 11;
}
```

From N3650:

 A resumable function is a function that is capable of split-phase execution, meaning that the function may be observed to return from an invocation without producing its final logical result or all of its sideeffects.

This allows

- Writing asynchronous code as if it was synchronous
- Avoids creating stack frames as the resumable function always leaves the stack in 'virgin' state

http://www.open-std.org/jtc1/sc22/wg21/docs/papers/2013/n3650.pdf

Disadvantage: requires compiler support

- Local variables and parameters have to be placed in heap allocated memory
- Certain transformations have to be applied to function bodies to create current continuations and to allow for re-entrance
- Heap based allocations necessary to store local variables and parameters
- Surprising semantics (return type, side effects of surrounding code, etc.)

Advantages

- Only one stack segment is required for each OS-thread
- Simplified code
- More asynchrony as functions can proceed whenever a called function awaits

Parallel way (6th attempt), using resumable functions

```
future<uint64 t> fibonacci(uint64 t n) async
    if (n < 2) return make_ready_future(n);</pre>
    if (n < threshold) return make_ready_future(fibonacci_serial(n));</pre>
    future<uint64_t> f = async(launch::async, &fibonacci, n-1); // .unwrap()
    future<uint64 t> r = fibonacci(n-2);
    return await f + await r;
```

Fastest parallelization results, outperforms all others by a significant amount

- Scales better
- Runs faster
- Uses less memory (less stack segments)
- Creates less threads

Resumable functions are a valuable addition to the language

- Simplify code, makes asynchronous code look synchronous
- Required code transformation are almost trivial and well understood

However, resumable function alone are not sufficient, they work best on code which has already been parallelized

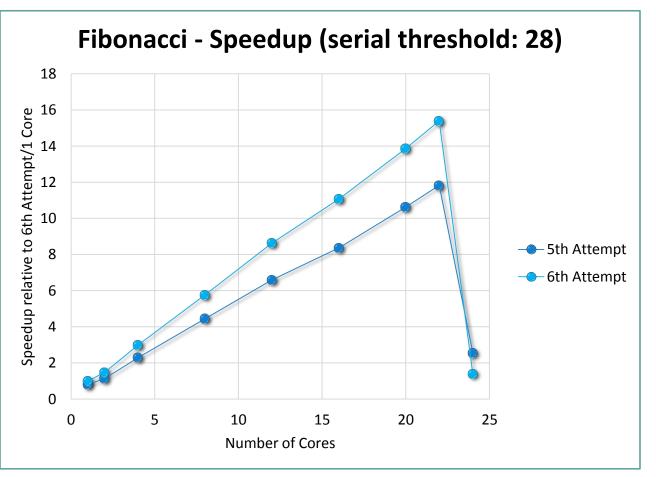
Fastest parallelization results, outperf

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- Creates less threads

Resumable functions are a valuable ac

- Simplify code, makes asynchronous cod
- Required code transformation are almore

However, resumable function alone a been parallelized



Parallel way (7th attempt), using resumable functions

```
future<uint64 t> fibonacci(uint64 t n)
    if (n < 2) return make ready future(n);</pre>
    if (n < threshold) return make ready future(fibonacci serial(n));</pre>
    future<uint64 t> f = async(launch::async, &fibonacci, n-1);
    future<uint64 t> r = fibonacci(n-2);
    return dataflow(
        [](future<uint64 t> f1, future<uint64 t> f2) -> uint64 t {
            return f1.get() + f2.get();
        f, r);
```

Thanks to Thomas Heller for coming up with this idea

So What's the Deal?

Too much parallelism is as bad as is too little

Sweetspot is determined by the Four Horsemen, mainly by contention

Granularity control is crucial

- Optimal grain size depends very little on number of used resources
- Optimal grain size is determined by the Four Horsemen, mainly by overheads, starvation, and latencies

Even problems with (very) strong data dependencies can benefit from parallelization

Doing more is not always bad

- While we added more overheads by futurizing the code, we still gained performance
- This is a result of the complex interplay of starvation, contention and overheads in modern hardware

Avoid explicit suspension as much as possible, prefer continuation style execution flow

Dataflow style programming is key to managing asynchrony

Predicting the Future

THE UNIVERSAL SCALABILITY LAW

Universal Scalability Law

$$S(N) = \frac{N}{1+\alpha (N-1)+\beta N(N-1)}$$

S(N) – Normalized scaling

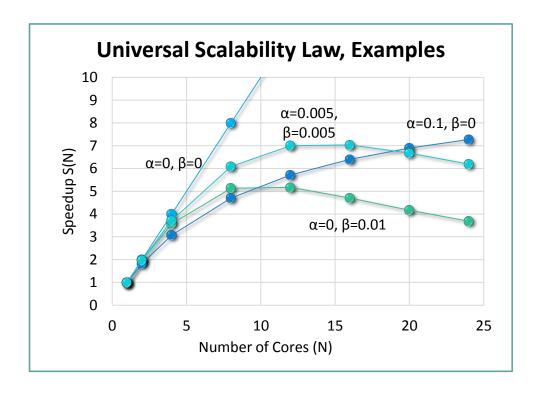
- 1 Concurrency
- Without interaction the function would scale linearly

 α (N - 1) – Contention, (time spent in scalar mode, i.e. α is serial fraction), starvation

 $^{\circ}$ Represents the degree of serialization on shared writable data and is parameterized by the constant σ .

 β N(N - 1) – Latency (Point-to-point), overheads

 Represents the penalty incurred for maintaining consistency of shared writable data



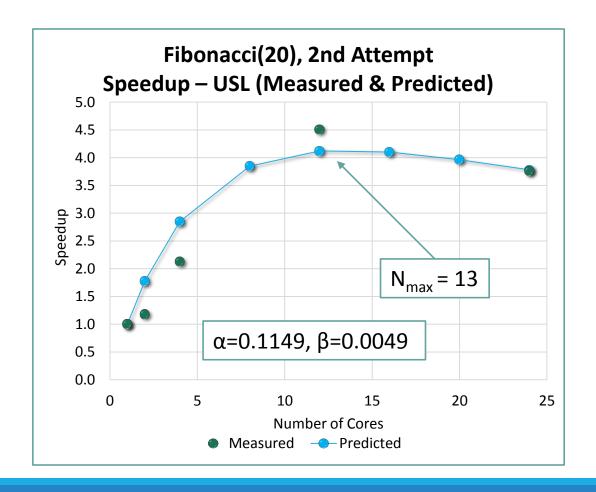
Universal Scalability Law

If we know the coefficients (α, β) we can estimate the N gaining the best speedup:

$$N_{max} = \sqrt{\frac{1-\alpha}{\beta}}$$

This does not tell us what causes the problems, but helps optimizing resource utilization

Can be done at runtime, 4-6 data points are sufficient



Using HPX

EXTENDING THE STANDARD BEYOND ONE NODE

HPX feature(s) for Asynchronous Computing:

Actions

- Actions are essentially functions in conventional sense
- But additionally can be invoked to remote locations

Global Actions

Functions that could to be invoked remotely.

```
void some_global_function(double d)
{
    cout << d;
}

// This will define the action type 'some_global_action' which represents
// the function 'some_global_function'.

HPX_PLAIN_ACTION(some_global_function, some_global_action);</pre>
```

Example

```
// Evaluating Factorial
boost::uint64_t factorial(boost::uint64_t x);
HPX_PLAIN_ACTION(factorial, factorial_action);
boost::uint64_t factorial(boost::uint64_t n)
    if (n <= 0) return 1;
    factorial_action fact;
    hpx::future<boost::uint64 t> n1 =
        hpx::async(fact, hpx::find_here(), n - 1);
    return n * n1.get();
```

Asynchronous Actions without Synchronization

- Fully asynchronous, controlling thread does not wait for action to start or complete
- Return value ignored if the function had return value

```
some_global_action act;  // define an instance of some_global_action
hpx::apply(act, hpx::find_here(), data);
```

```
some_component_action act;  // define an instance of some_component_action
hpx::apply(act, id, "42");  // id is component's global id
```

Asynchronous Actions with Synchronization

- Controlling thread does not wait for the function to start or complete
- Waits for the return value of the function

```
some_global_action act;  // define an instance of some_global_action
hpx::future<void> f = hpx::async(act, hpx::find_here(), 2.0);

//
// ... do other stuff here
//
f.get();  // this will possibly wait for the asyncronous operation to 'return'
```

Synchronous Actions

- The invoked function is scheduled immediately
- The calling thread waits for the function to complete

```
some_global_action act;  // define an instance of some_global_action
act(hpx::find_here(), 2.0);
```

```
some_component_action act;  // define an instance of some_component_action
int result = act(id, "42");
```

Asynchronous Actions With Continuation But Synchronization

- Similar to async action with synchronization; but takes additional function argument.
- Similar to future function().then(....) on the same locality

```
// first action
boost::int32_t action1(boost::int32_t i)
{
    return i+1;
}
HPX_PLAIN_ACTION(action1); // defines action1_type

// second action
boost::int32_t action2(boost::int32_t i)
{
    return i*2;
}
HPX_PLAIN_ACTION(action2); // defines action2_type
```

```
action1_type act1;  // define an instance of 'action1_type'
action2_type act2;  // define an instance of 'action2_type'
// action1_type and action2_type are global function or component function.
hpx::future<int> f =
    hpx::async_continue(act1, hpx::find_here(), 5,
        hpx::make_continuation(act2));
hpx::cout << f.get() << "\n";  // will print: 12 ((5+1) * 2)</pre>
```

Continuing to remote locality

The final return of chain of called function goes back to origin

Chaining more than 2 operations

Asynchronous Actions With Continuation But Without Synchronization

- Similar to async actions with no synchronization,
- Similar structure as above, but after evaluation of the last function, the return value if any is discarded

Components and Actions (Distributed Computing – HPX)

Components

- "Remotable" C++ objects
- These are First Class objects in HPX, with a globally unique name, or GID
- Context of the instantiated object is preserved, for the duration of the client side that holds reference to its global name.

Component Actions

Functions that need to be invoked remotely, but are part of a class object/ component object.

Components in HPX

```
class accumulator
  : public hpx::components::simple component base<accumulator>
public:
    accumulator() : acculumlate_(0) {}
   void add(double to_add) { accumulate_ += to_add; }
    double query() const { return accumulator_; }
   HPX_DEFINE_COMPONENT_ACTION(accumulator, add, add_action);
   HPX DEFINE COMPONENT CONST ACTION(accumulator, query, query action);
private:
    double accumulate;
};
```

Components in HPX

```
// create new instance of an accumulator
hpx::id type id = hpx::new_<accumulator>();
accumulator_add_action add_action;
// Explicitly asynchronous invocation
hpx::future<void> f = hpx::async(add_action, id, 42.0);
// ...
f.get();
// Explicitly 'synchronous' invocation
accumulator_query_action query_action;
hpx::cout << query_action(id) << "\n"; // prints: 42</pre>
// fire and forget version
hpx::apply(add action, id, 43);
```

The Life of Pi

Embarrassingly Parallel Applications

Easily broken into approximately equal amounts of work per processors

Each Individual Task (per processor) is independent, i.e. minimal communication among processors

We get near-perfect parallel speedup with modest programming efforts

M work load can be divided into N chunk, with each work load equaling M/N

Parallel Overhead due to communication (over work coordination) and/or reduce operation at the end

Examples:

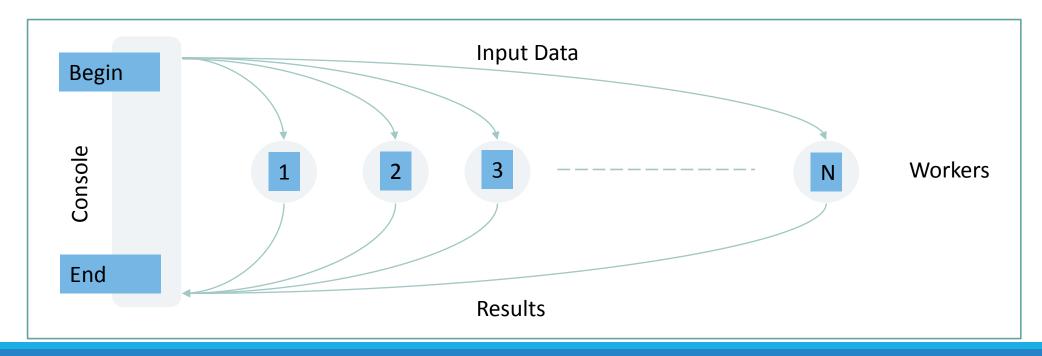
- Image Processing,
- Monte Carlo Simulation
- Random Number Generation
- Encryption, Compression

Embarrassingly Parallel Application

$$Y_i = F_i(X_i)$$

```
, where i = 0, 1, 2, ..., N.
```

, x_i -> inputs, y_i -> outputs and, F_i -> pure function



Monte Carlo

Monte Carlo Methods: simulating physical phenomena based on randomness

- Randomly generate a large number of example cases (input) of a phenomenon,
- Perform some computation on these inputs
- Take average of the observed results

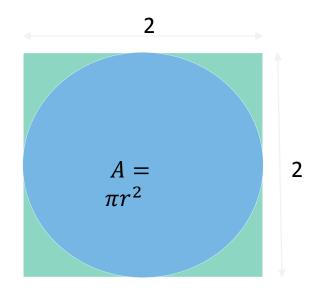
The result is an approximation to some true but unknown quantity

Monte Carlo Simulations are typically embarrassingly parallel

Each unit simulation is completely independent from all such unit simulations.

Evaluation of Pi (Monte Carlo)

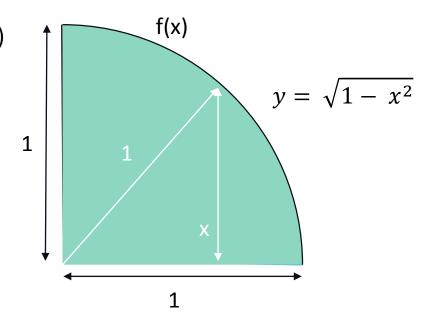
Area of Circle
Area of Square =
$$\frac{\pi r^2}{2 \times 2} = \frac{\pi}{4}$$



Random Pairs of Numbers, (x_r, y_r)

$$x_r^2 + y_r^2 \le 1$$

$$\int_0^1 \sqrt{1 - x^2} \, dx = \frac{\pi}{4}$$



Area =
$$\int_{x_1}^{x_2} f(x) dx = \lim_{N \to \infty} \frac{1}{N-1} \sum_{i=0}^{N-1} f(x_i) (x_2 - x_1)$$

Evaluation of Pi using HPX

```
// define action: can be called remotely
double pi_montecarlo_evaluate(std::size_t num_iteration)
   // ...
// defines pi_montecarlo_evaluate_action
HPX PLAIN ACTION(pi montecarlo evaluate);
// retrieve all participating localities (nodes)
std::vector<hpx::id type> localities =
     hpx::find all localities();
```

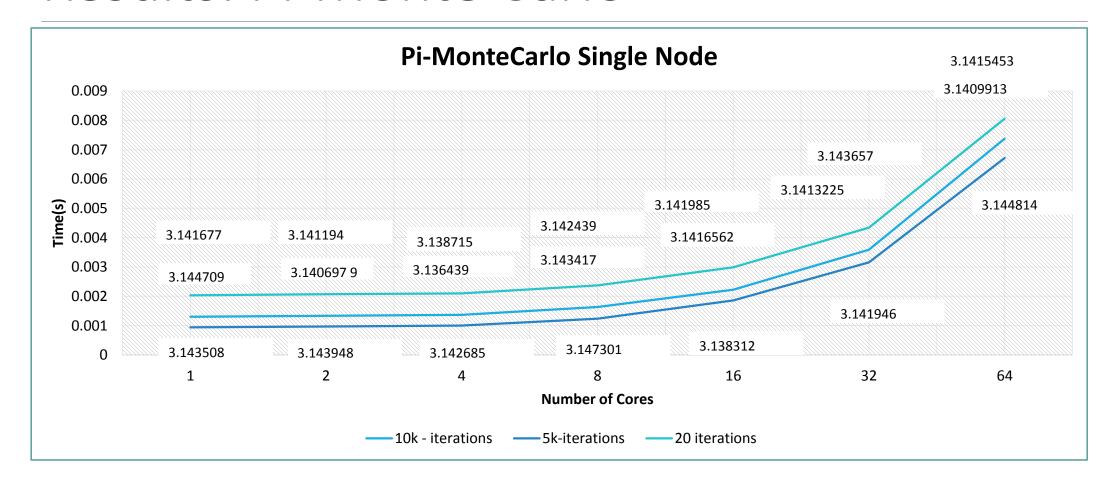
Evaluation of Pi using HPX

```
double pi montecarlo evaluate(std::size t num iteration)
    base_generator_type generator(0.1);
    generator.seed(std::time(0));
    gen_type monte_carlo(generator, distribution_type(0, 1));
    boost::generator iterator<gen type> gen value(&monte carlo);
    double x, area = 0;
    // dx = 1 / num_iteration;
    for(std::size t i = 0; i != num iteration; ++i)
        x = *gen value++;
         area += std::sqrt(1 - x*x); // f(x) * dx
    // return area of quadrant
    return area / num iteration;
```

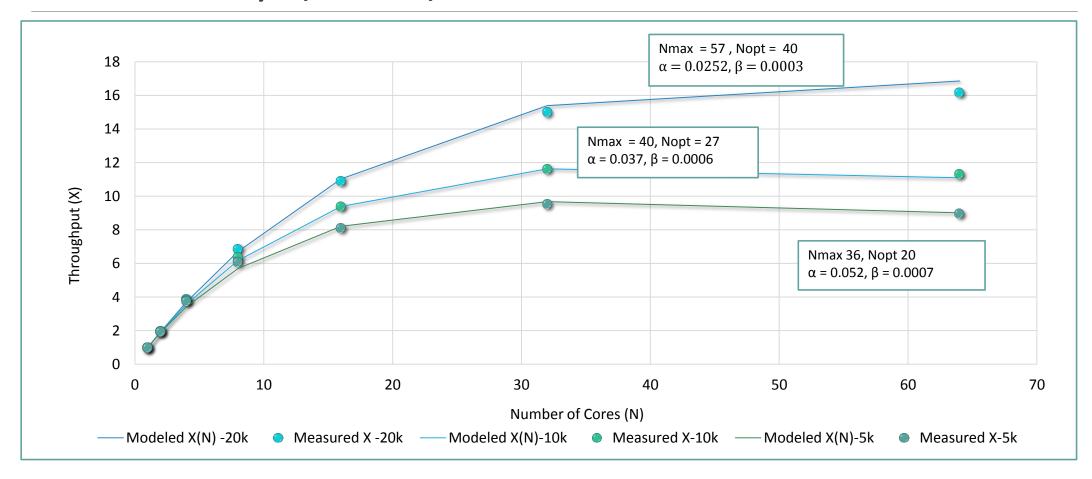
Evaluation of Pi using HPX

```
std::vector<future<double>> futures;
pi_montecarlo_evaluate_action act;
BOOST_FOREACH(id_type const& node, localities)
    futures.push_back(async(act, node, num_iter));
wait all(futures);
double result = 0, pi = 0;
BOOST_FOREACH(future<double>& fut, futures)
    result += fut.get();
pi = 4 * result;
```

Results: Pi-Monte Carlo



Results: Pi – Monte Carlo USL projected Scalability (SMP)



Fast Fourier Transform (FFT)

Well Known and Understood problem

Data dependency between elements is well defined

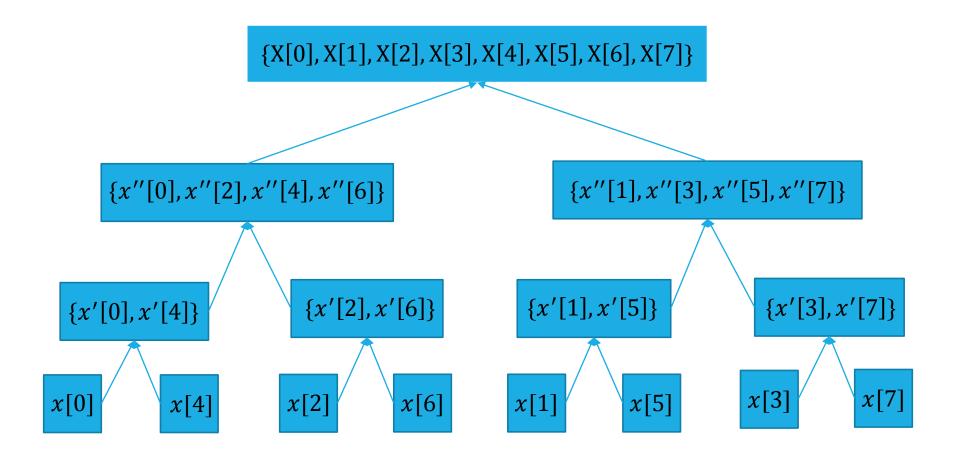
Parallel Implementation of FFT Emerging recently with several parallelization techniques available

- In Distributed FFT, data explicitly need to be moved
 - So a parallel asynchronous computation mechanism would be nice.
 - HPX async

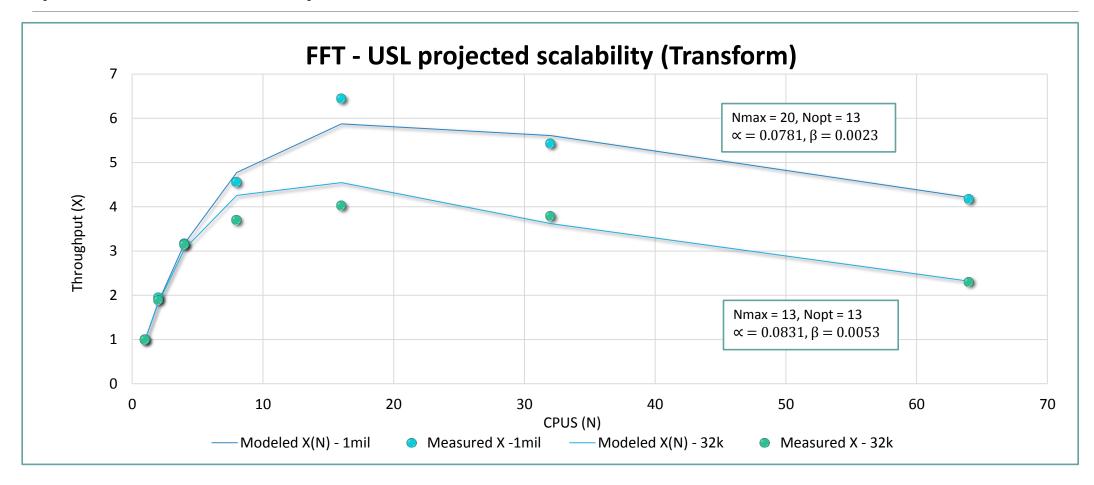
Implementation of Parallel FFT in HPX

Based on Coley-Tuckey algorithm with Radix – 2DIT

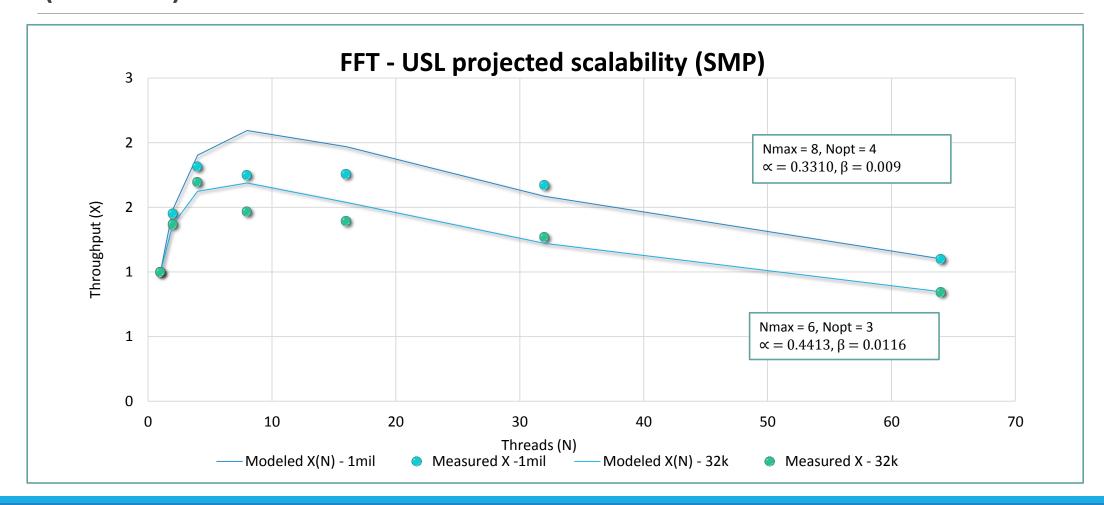
FFT (Decimation in Time)



Results: FFT - USL Projected Scalability (Transform)



Results: FFT - USL Projected Scalability (SMP)



Conclusions

Conclusions

Be aware of the Four Horsemen

Embrace parallelism, it's here to stay, avoid concurrency

Asynchrony is your friend if used correctly

Think in terms of data dependencies, make them explicit

Avoid thinking in terms of threads

Continuation style, dataflow based programming is key for successful parallelization

Performance modelling can help adjusting parameters

Where to get HPX

Main repository: https://github.com/STEllAR-GROUP/hpx/ (Boost licensed)

Main website: http://stellar.cct.lsu.edu/

Mailing lists: hpx-users@stellar.cct.lsu.edu, hpx-devel@stellar.cct.lsu.edu

IRC channel: #ste | | ar on freenode