Presented by: Jonathan Beard To: C++Now 2016

Alternate Titles:

This thing I started on a plane

Alternate Titles:

This thing I started on a plane What's this RaftLib Thingy?

### Alternate Titles:

### This thing I started on a plane What's this RaftLib Thingy? OMG, Another Threading Library...

### Alternate Titles:

This thing I started on a plane What's this RaftLib Thingy? OMG, Another Threading Library... Why I hate parallel programming

Alternate Titles:

This thing I started on a plane What's this RaftLib Thingy? OMG, Another Threading Library... Why I hate parallel programming A self help guide for pthread anxiety



All thoughts, opinions are my own. RaftLib is not a product of ARM Inc. Please don't ask about ARM products or strategy. I will scowl and not answer.

Thank you 😎

## **ABOUT ME**

my website

http://www.jonathanbeard.io



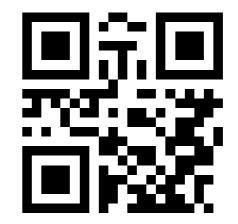
slides at

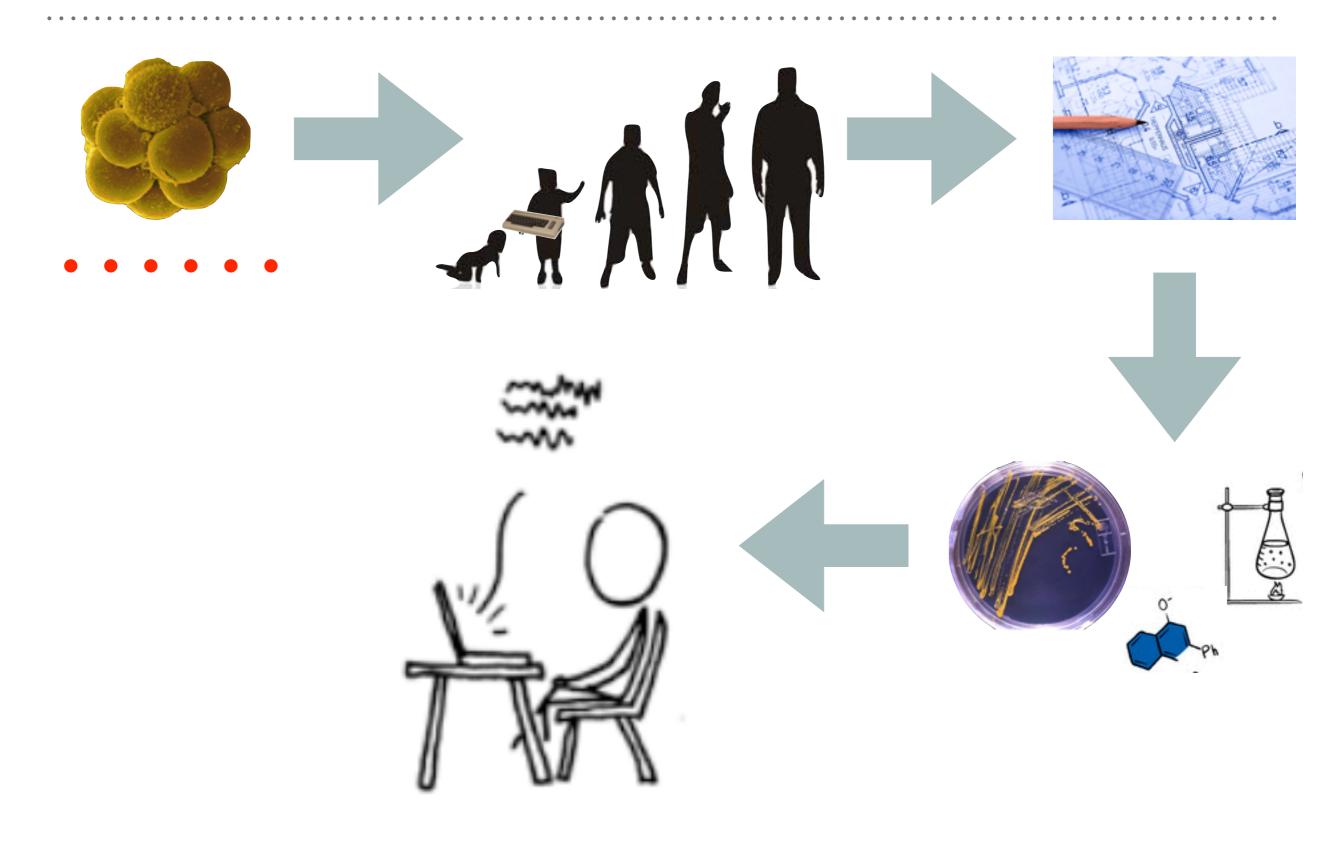
http://goo.gl/cwT5UB

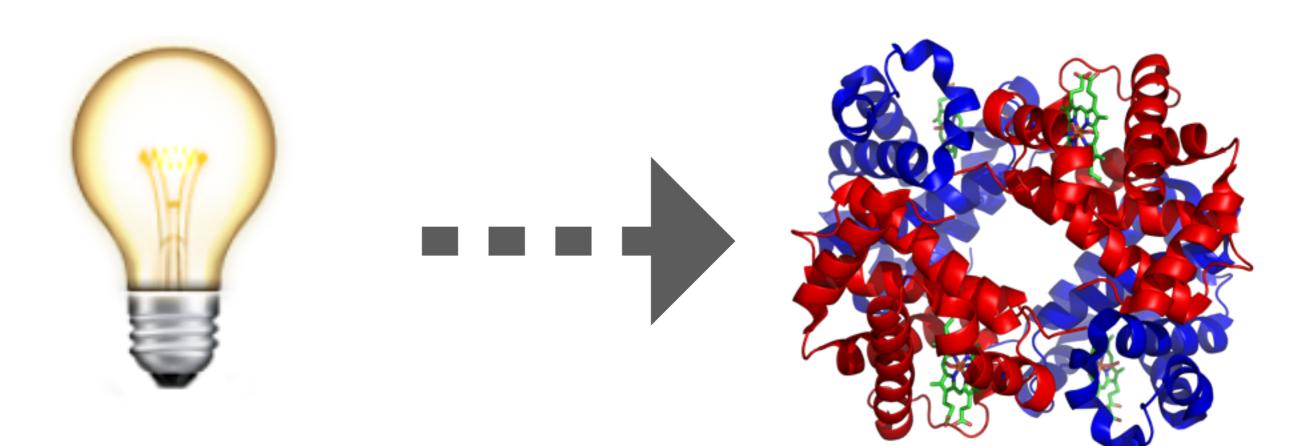


project page

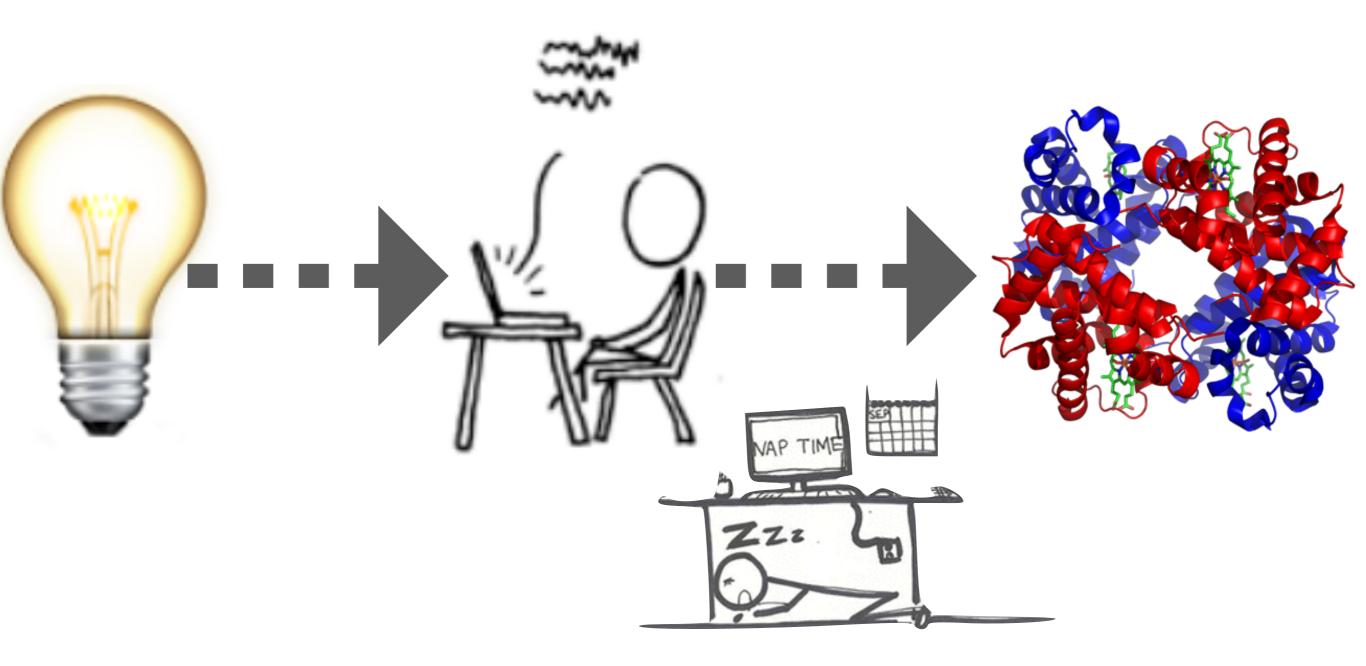
<u>raftlib.io</u>

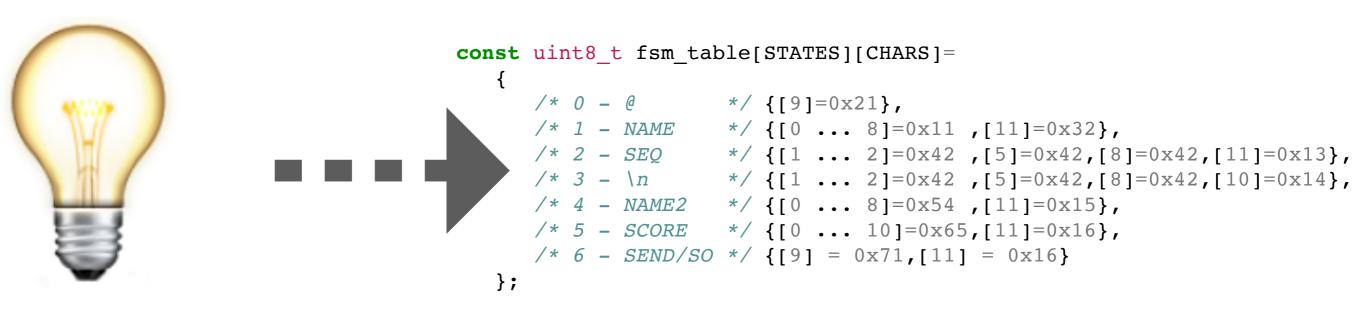




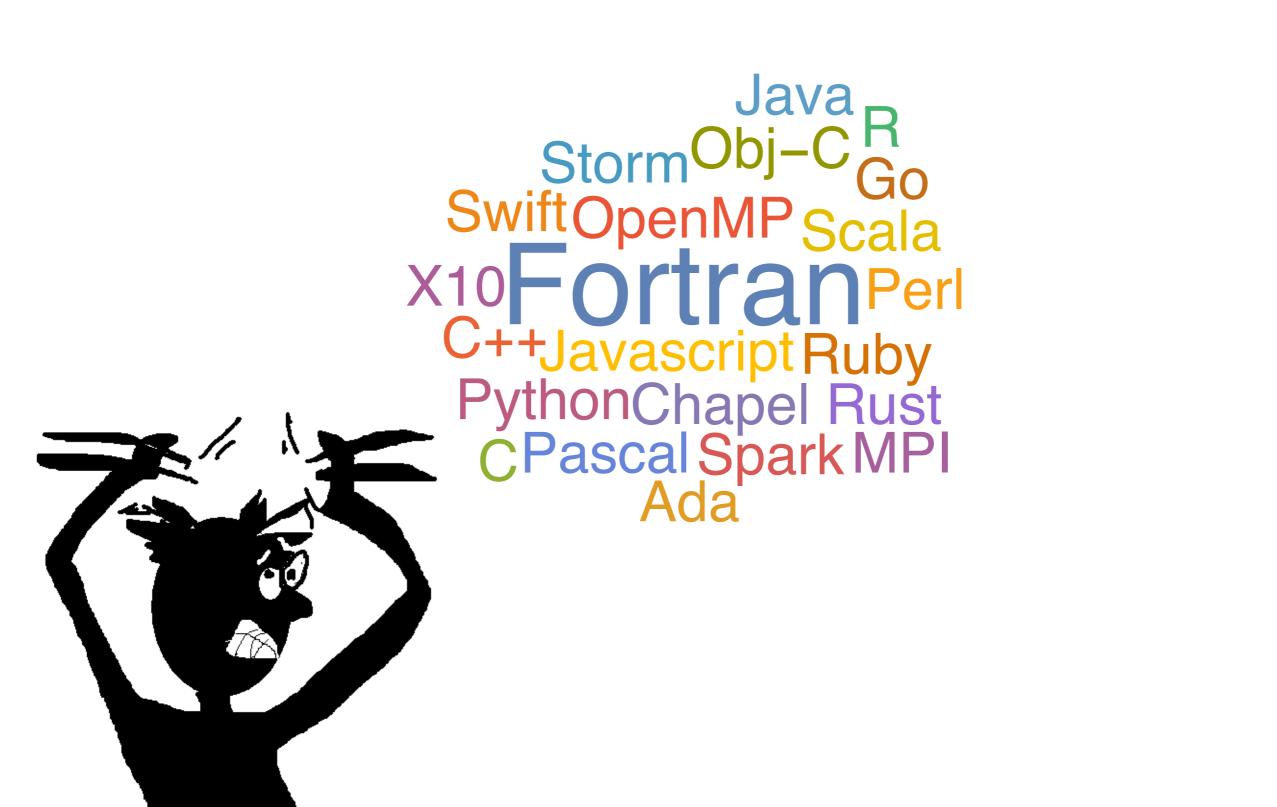


. . . . . . . . . . . . . . . . . .

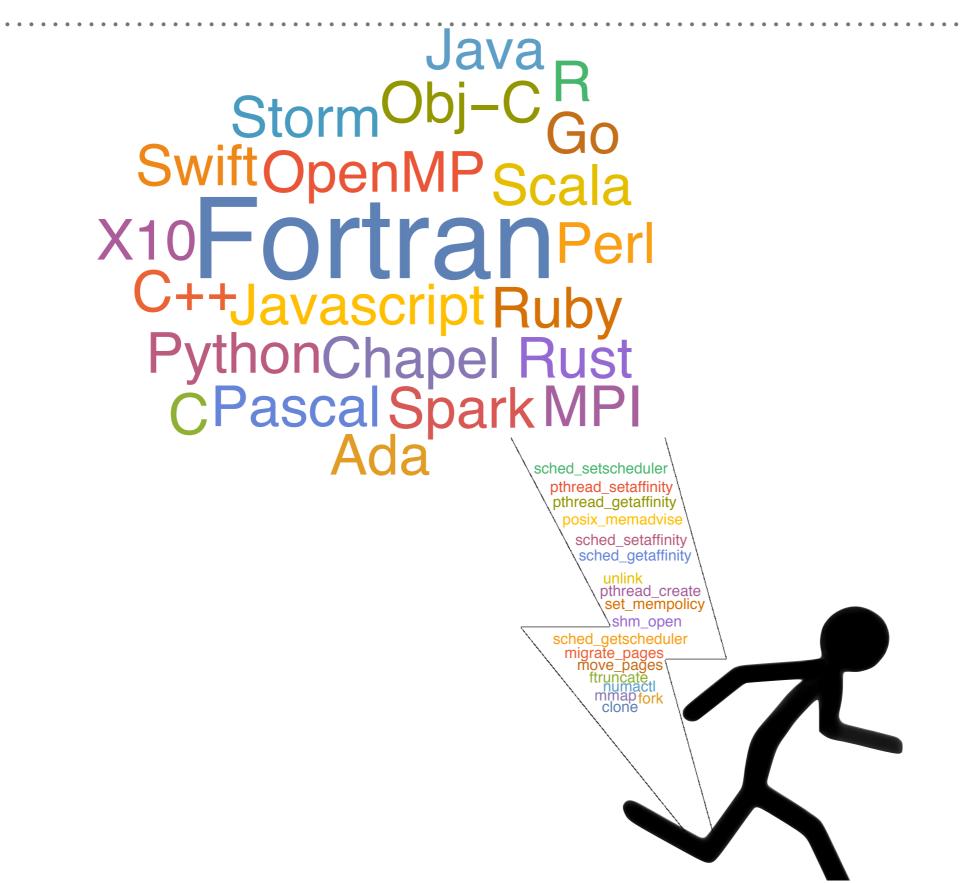




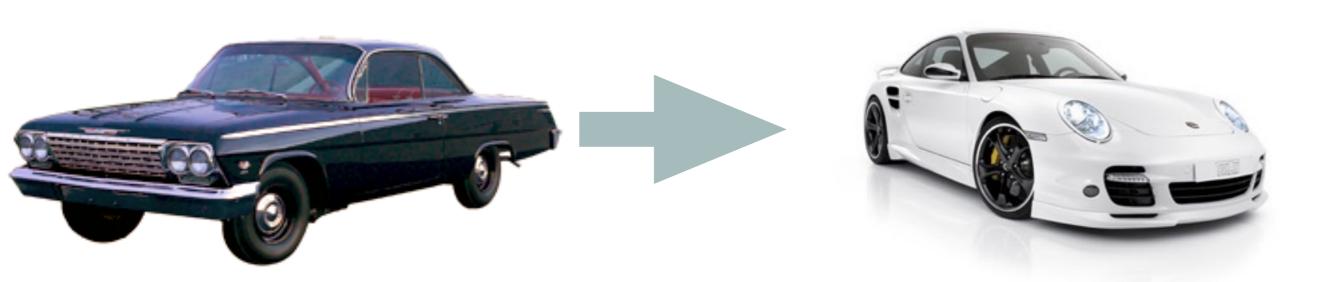
## **NECESSITY DRIVES IDEAS**



## **NECESSITY DRIVES IDEAS**

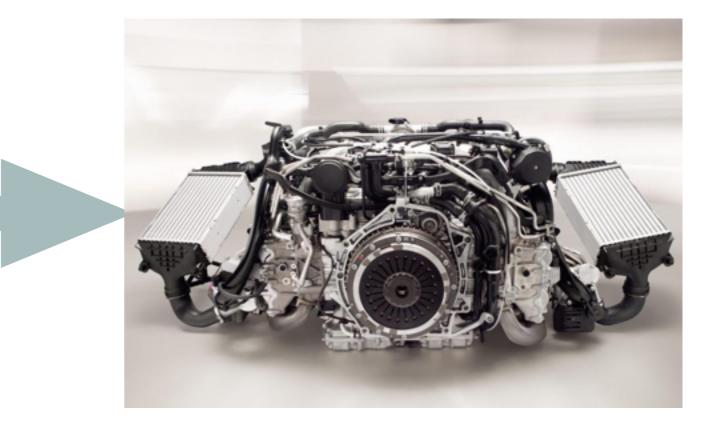


## AN (PERHAPS BAD) ANALOGY



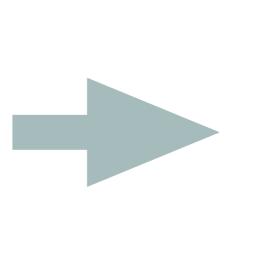
## AN (PERHAPS BAD) ANALOGY





## AN (PERHAPS BAD) ANALOGY





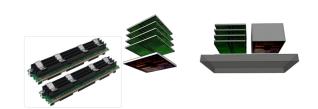


## ANALOGY PART TWO





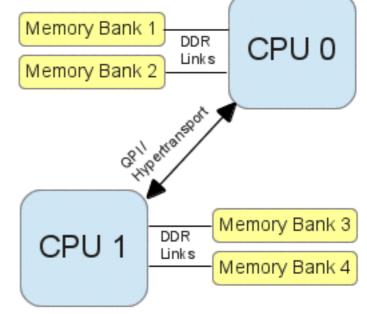




## TOPOLOGY

Uniform Memory Access Frontside CPU 1 CPU 0 Bus **Memory Controller** HUB (Northbridge) DDR Links CPU 1 Memory Bank 3 Memory Bank 1 Memory Bank 4 Memory Bank 2 300 <sub>F</sub> 250 Nanoseconds 200 150 100 50 0 Different Same

### Non-Uniform Memory Access



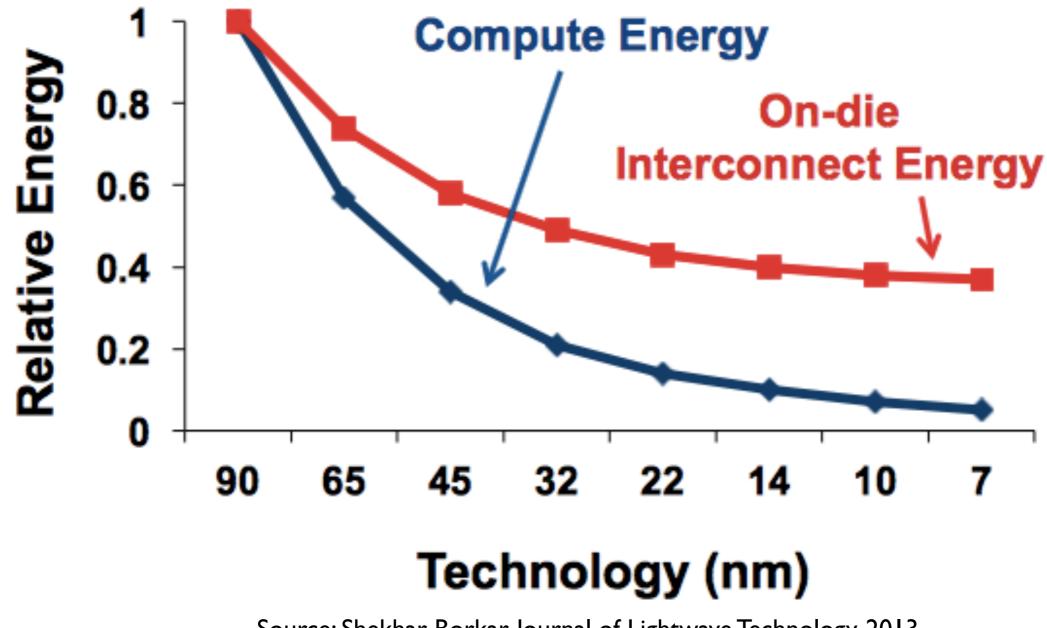
## THE JELLYFISH



## THE FIRST ORGANISM TO OVERLAP ACCESS AND EXECUTION

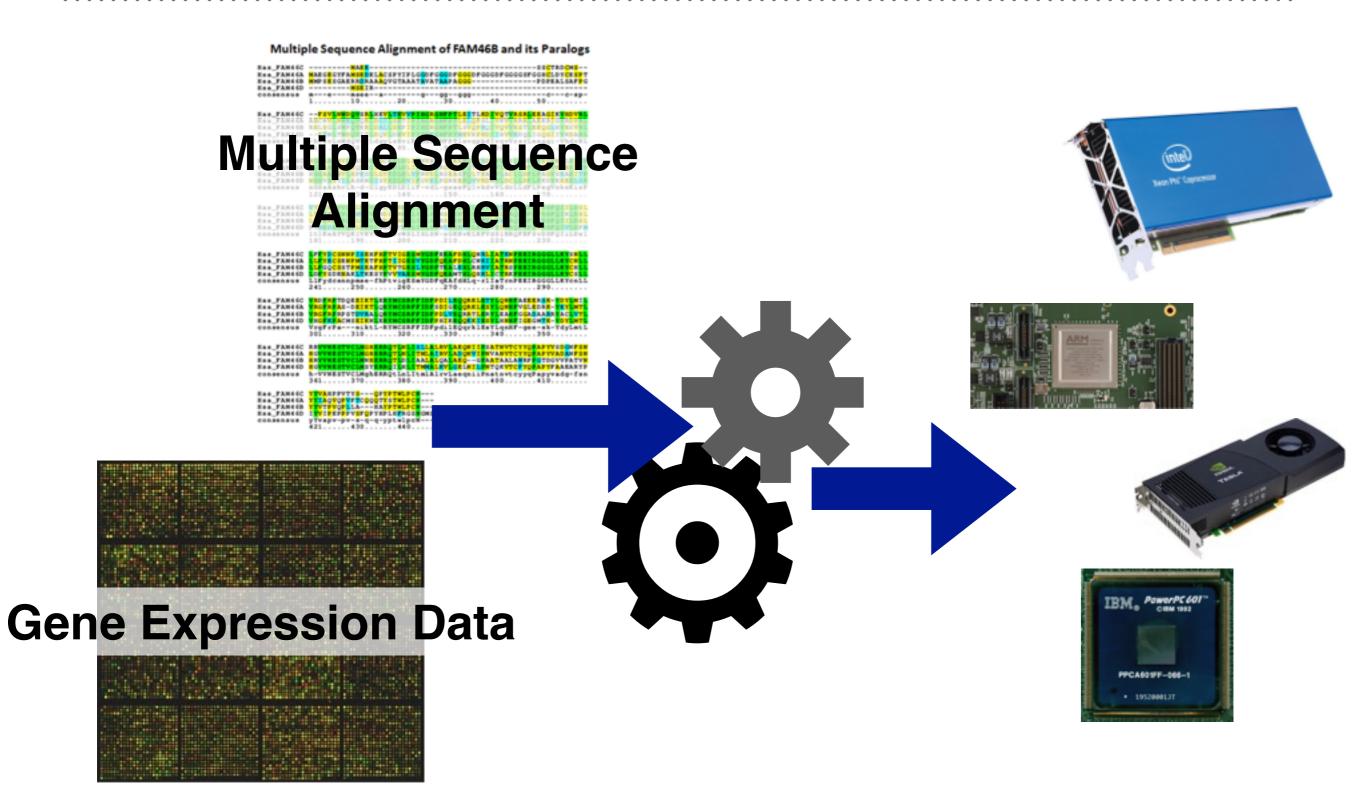


## DATA MOVEMENT DOMINATES

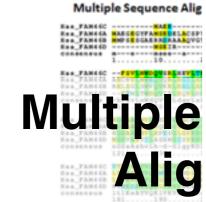


Source: Shekhar Borkar, Journal of Lightwave Technology, 2013

## I SHOULDN'T HAVE TO CARE



## I SHOULDN'T HAV



. .

 Ram
 FAN44C
 FFFCCMMP
 FEE

 Ram
 FAN44A
 FFECCMMP
 FFE

 Ram
 FAN44A
 FFECCMMP
 FFE

 Ram
 FAN44A
 FFECCMMP
 FFE

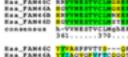
 Conserve
 SIAP
 FFECCMP
 FFE

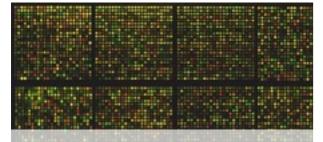
 Conserve
 SIAP
 FFE
 FFE

 Zeit
 Zeit
 Zeit
 Zeit
 Zeit

 Ram
 FAN44C
 WDUPNETOURS
 Zeit
 Zeit
 Zeit

Ess_FAM46A	VROFREAS-DELETED
	VECTATEPSTOVEALO
Ess_FAM46D	A ROLE ACHIELEN E
COR868848	VrgFrFaelkt1- 301
EAS TANGE	A REPORT OF LOCAL DISTANCE IN CAMERA DISTANCE INCORDA DISTANCE IN CAMERA DISTANCE INCORDA DISTANCE IN CAMERA DISTANCE INCORDA DISTANCE DISTANCE DISTANCE DISTANCE DISTANCE DISTANCE DISTANCE INCORDA DISTANCE DISTANC











. . . . . . . . . . . . .

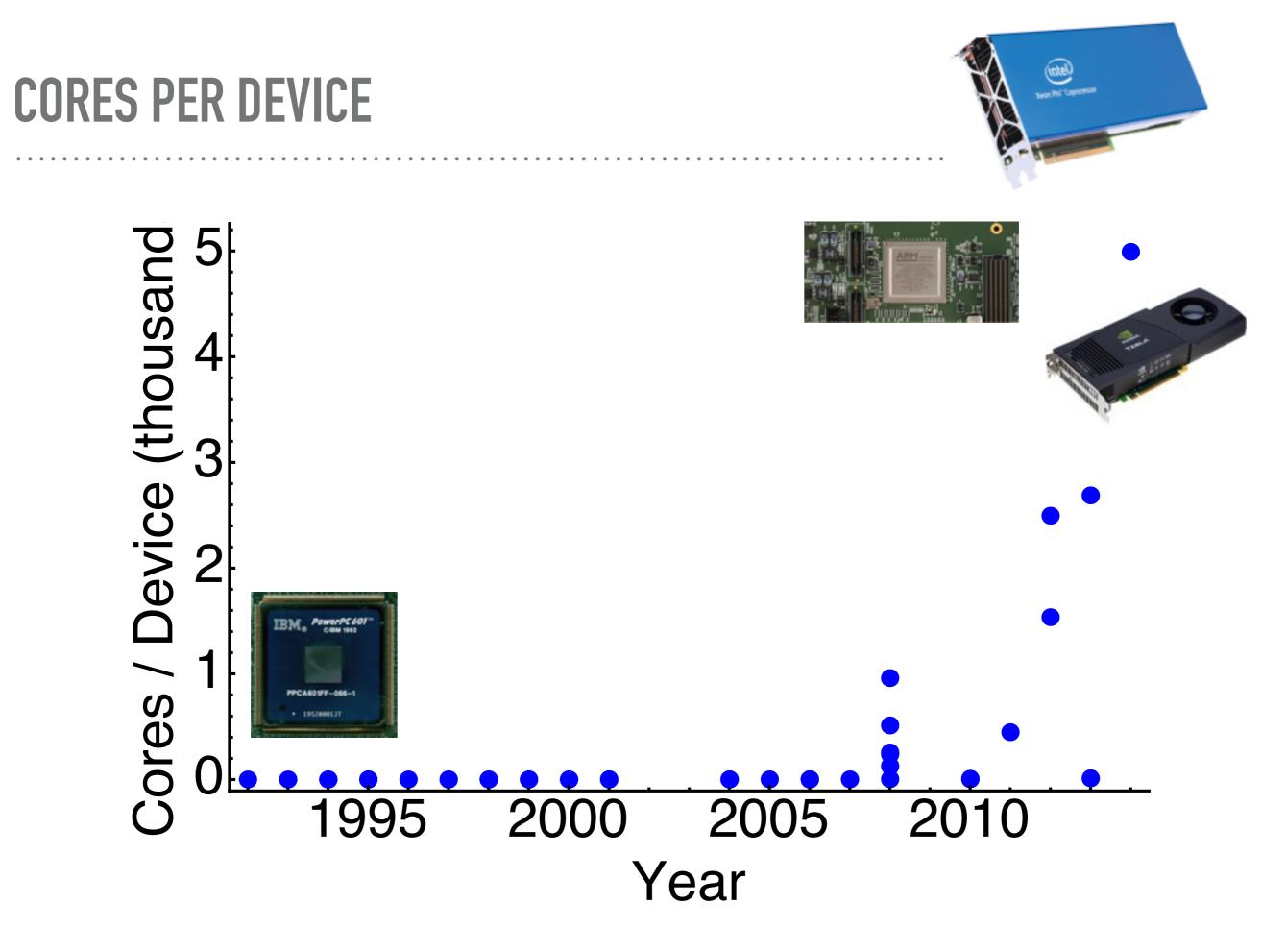
## HARDWARE

### **1984** Cray X-MP/48 \$19 million / GFLOP



**2015** \$.08 / GFLOP



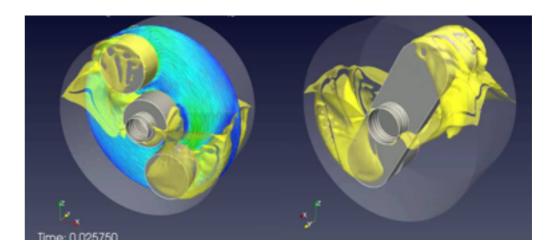


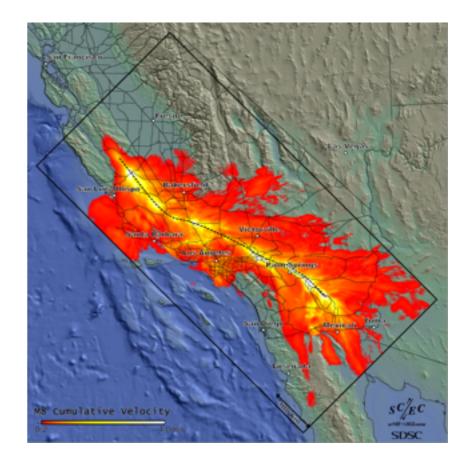
## FINANCIAL INCENTIVE

Sequential JS: \$4-7/line Sequential Java: \$5-10/line Embedded Code: \$30-50/line HPC Code: \$100/line

## WHY YOU SHOULD CARE







## PRODUCTIVITY / EFFICIENCY AN EQUALIZER

- Titan SC estimates for porting range from 5, to > 20 million USD
- Most code never really optimized for machine topology, wasting \$\$ (time/product) and energy
- ► Getting the most out of what you have is an equalizer





## HAVES AND HAVE NOTS

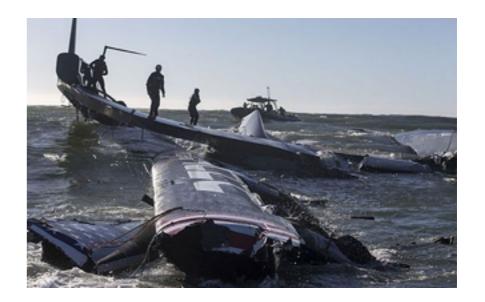
### **Gov't / Big Business**

- Lots of \$\$
- Can hire the best people
- Can acquire the rest
- Plenty of compute resources



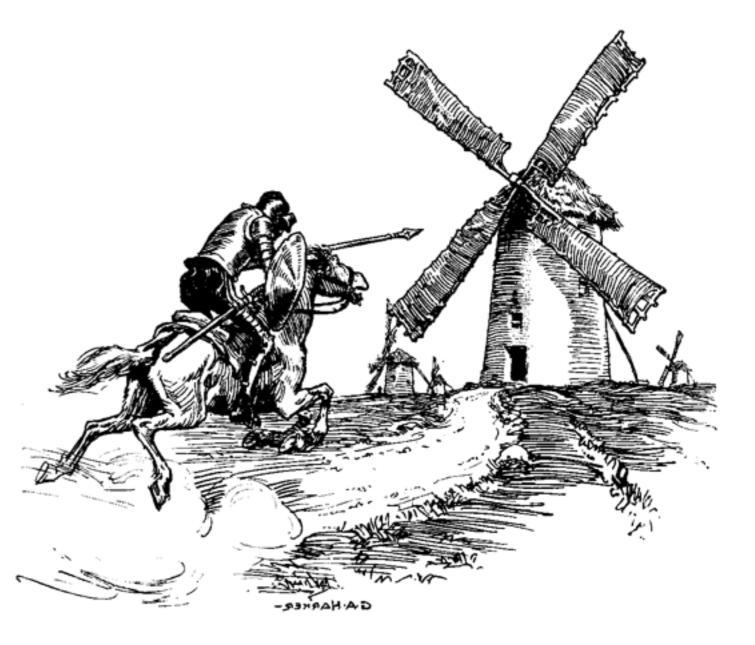
## Start-up / Small Government

- Not a lot of \$\$
- Often can't hire the best people
- Left to the mercy of cloud providers





## Let's make computers super fast, and easy to program



## **WHERE TO START**

### Brook for GPUs: Stream Computing on Graphics Hardware

### Ian Buck Tim Foley Daniel Horn Jeremy Sugerman Kayvon Fatahalian Mike Houston Pat Hanrahan

Stanford University

### Abstract

In this paper, we present Brook for GPUs, a system for general-purpose computation on programmable graphics hardware. Brook extends C to include simple data-parallel constructs, enabling the use of the GPU as a streaming co-processor. We present a compiler and numine system that abstracts and virtualizes many aspects of graphics hardware. OPIII as a computer genine accomposite to the GPU is a streaming optimised on the strength of the stream of In addition, we present an analysis of the effectiveness of the  ${\rm FeU}$  as a compute engine compared to the CPU, to deter-mine when the GPU can outperform the CPU for a particu-tion of the CPU can outperform the CPU for a particu-tion of the SADY and SCHAM BLAS coperators, image segmen-tation, FPT, and ray tracing. For these applications, we demonstrate that our Brook implementations perform com-parably to hand-written GPU code and up to seven times faster than their CPU counterparts. CR Categories: L3.1 [Computer Graphics]: Hard-ware Architecture—Graphics processors D.3.2 [Program-ming Languages]: Language Classifications—Parallel Lan-guages

Keywords: Programmable Graphics Hardware, Data Parallel Computing, Stream Computing, GPU Computing,

tion and optimization, we can scale applications to a large number of nodes. Currently, we can run SPADE jobs on  $\approx 500$  processors within more than 100 physical nodes in a tightly connected cluster environment. SPADE has been in use at IBM Research to create real-world streaming appli-

Permission to make digital or hard copies of all or part of this work for personal or classroom use is granted without fee provided that copies are not made or distributed for profit or commercial advantage and that copies bear this notice and the full citation on the first page. To copy otherwise, to republish, to post on servers or to redistribute to fusts, requires prior specific permission and/or a fee. SIGMOD'08, June 9–12, 2008, Vancouver, BC, Canada. Copyright 2008 ACM 978-1-60558-102-6/08/06 ...\$5.00.

modern hardware. In addition, the user is forced to espect ther algorithm in terms of graphics primitives, such as set setures and triangles. As a result, general-purpose GPU computing is limited to only the most advanced graphics developers. The user of the GPU as a streaming coprocessor. The main contributions of the paper are:
• The presentation of the Brook stream program EDEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING the used for general-purpose GPU computing. Three the user of streams kernels and production querates
• The presentation of the Brook stream program and production querates

model for general-purpose GPU computing. the use of streams, kernels and reduction of Brook abstracts the GPU as a streaming pro-

### The demonstration of how various GPU bardware lim-itations can be virtualized or extended using our com-piler and runtime system; specifically, the GPU mem-ory system, the number of supported **TOOLS.FOR SIMULATION AND ANALYSIS OF HETEROGENEOUS PIPELINED** and support for user-defined data structures.

ARCHITECTURES

The presentation of a cost model for comparing GPU vs. CPU performance tradeoffs to better understand under what circumstances the GPU outperforms the CPU.

Programmable graphics hardware dates back to the original programmable for the direction of Professor Mark A. Franklin nal programmable for the direction of t

<text><text><text><text><text><text><text><text><text><text><text> nating in the PoelFow machine [Mohan et al. 1976]. These systems embeddeds **Stream: Broccess**, the same state of the second MAKHERLOINGCHENCE

tributed data stream processing middleware under develop-ment at IBM T. J. Watson Research Center. As a front-end for rapid application development for System S, SPADE pro-vides (1) an intermediate language for flexible composition of parallel and distributed data-flow graphs, (2) a toolkit of type-generic, bull-in stream processing operators, that sup-port scalar as well as vectorized processing and can seam-lessly inter-operate with user-defined operators, and (3) a rich set of stream adapters to ingest/publish data from/to outside sources. More importantly, SPADE automatically brings performance optimization and scalability to System S applications. To that end, SPADE employs a code genera-tion framework to create highly-optimized applications that run natively on the Stream Processing Core (SPC), the exerun natively on the stream Processing Core (SFC), the exe-cution and communication substrate of System S, and take full advantage of other System S services. SPADE allows de-velopers to construct their applications with fine granular stream operators without worrying about the performance implications that might exist, even in a distributed system. SPADE's optimizing compiler automatically maps applica-tions into appropriately sized execution units in order to tions not appropriately start execution mins in order to minimize communication overhead, while at the same time exploiting available parallelism. By virtue of the scalability of the System S runtime and SPADE's effective code genera-

NY, 10532, USA Saure Center, Hawthorne, NY, 10532, USA Saure Cours, Jon Com

cations, ranging from monitoring financial market feeds to radio telescopes to semiconductor fabrication lines

### Categories and Subject Descriptors

H.2.4 [Database Management]: Systems—Distributed databases; H.2.3 [Database Management]: Languages— Data manipulation languages

General Terms

Design

### Keywords

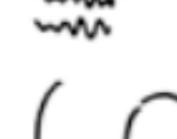
Distributed Data Stream Processing

### 1. INTRODUCTION

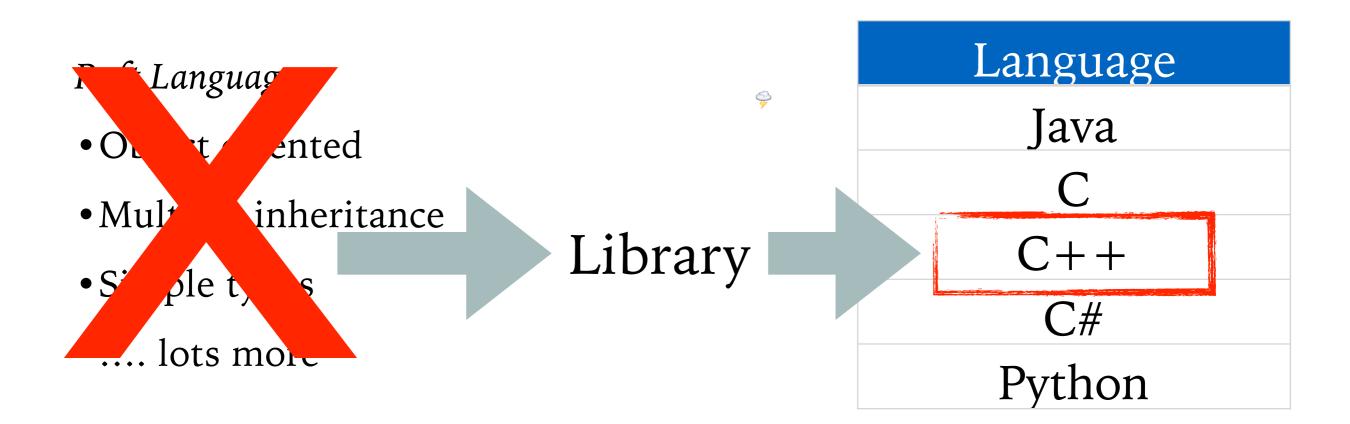
On-line information sources are increasingly taking the form of *data streams*, that is time ordered series of events or readings. Example data streams include live stock and or readings. Example data streams include live stock and option trading feeds in financial services, physical link statis-tics in networking and telecommunications, sensor readings in environmental monitoring and emergency response, and The proliferation of these sources has created a paradigm shift in how we process data, moving away from the traditional "store and then process" model of database manage-ment systems toward the "on-the-fly processing" model of emerging data stream processing systems (DSPS). This paradigm shift has recently created a strong interest in compared to the store of th DSPS-related research, in academia [1, 4, 6, 7] and industry [8, 17, 21, 25] alike. In this paper we describe the design of SPADE, which is

the declarative stream processing engine of the massively scalable and distributed System S - a large-scale stream scalable and distributed system S = a large-scale stream processing middleware under development at IBM Research. SPADE provides a rapid application development front-end for System S. Concretely, SPADE offers:

 An intermediate language for flexible composition of parallel and distributed data-flow graphs. This lan-guage sits in-between higher level programming tools and languages such as the System S IDE or stream



## WHERE TO START



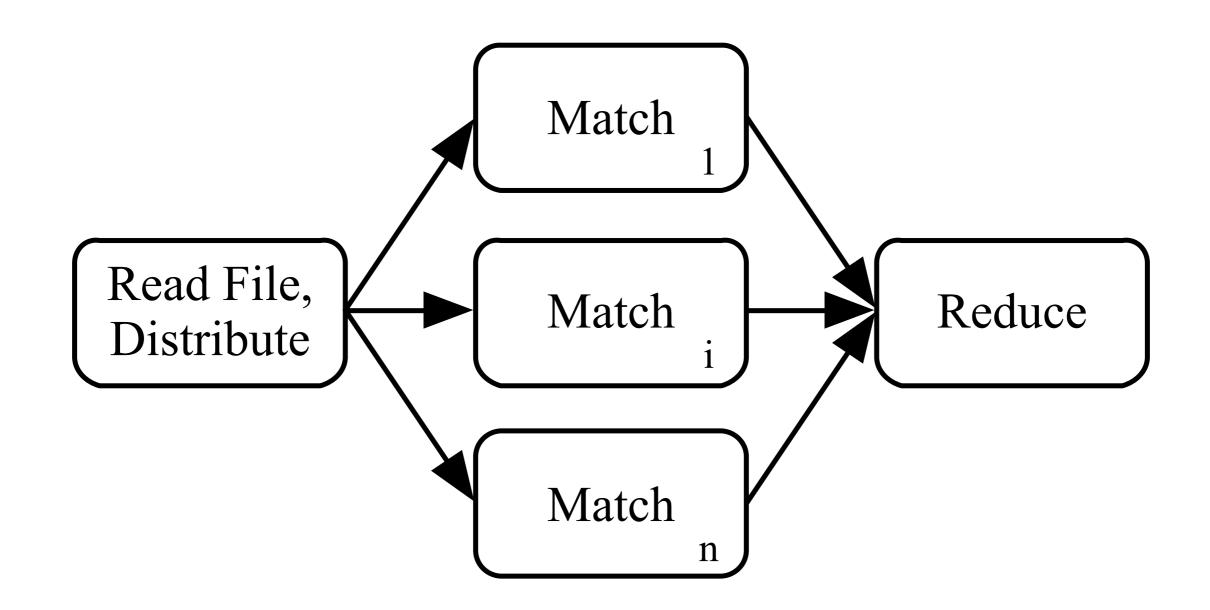
C++ Streaming Template Library Simplifies parallelization of code Abstracts the details Auto-manages allocation types, data movement.

software download: http://raftlib.io

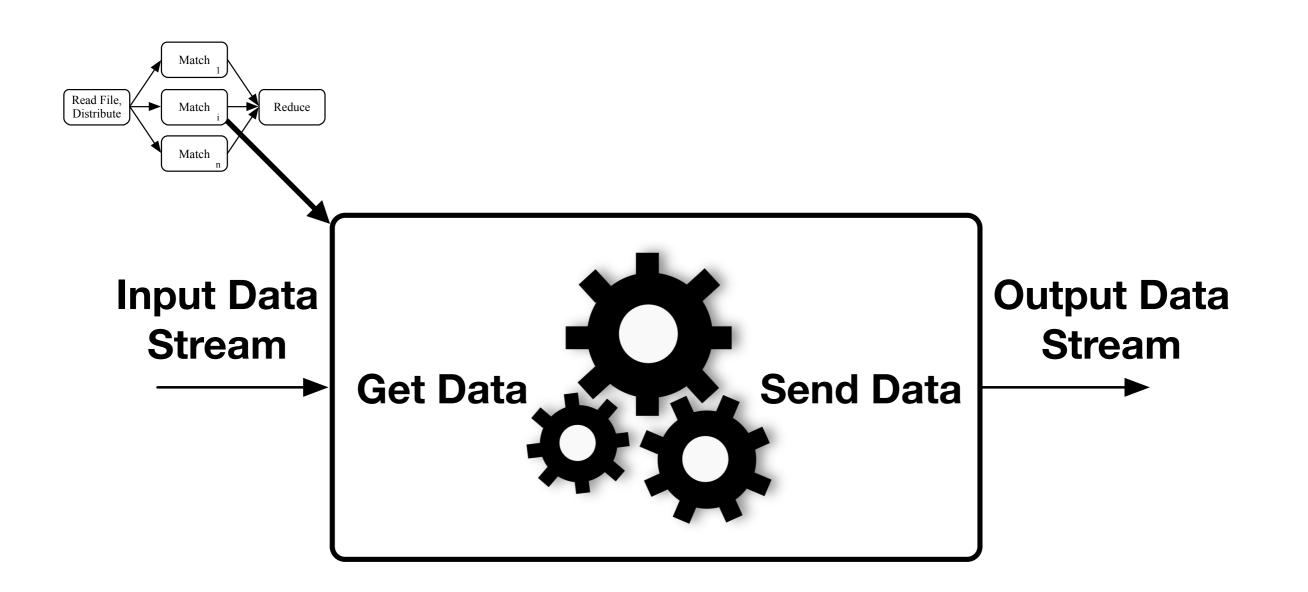
## **ISSUES THAT NEED SOLUTIONS**

- ► Usable Generic API
- ► Where to run the code
- ► Where to allocate memory
- ► How big of allocations to start off with
- ► How many user vs. kernel threads
- ► ...Gilligan had a better chance of having a 3 hour tour
- ...than we have of fitting all the issues on a single slide

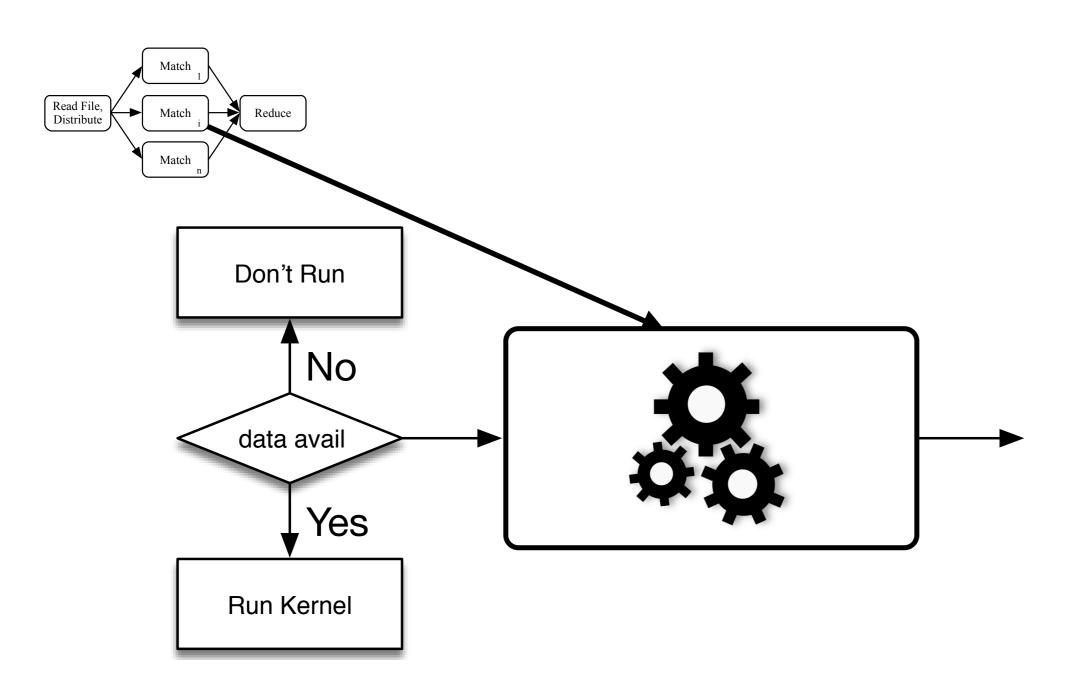
## STREAM PROCESSING (STRING SEARCH EXAMPLE)



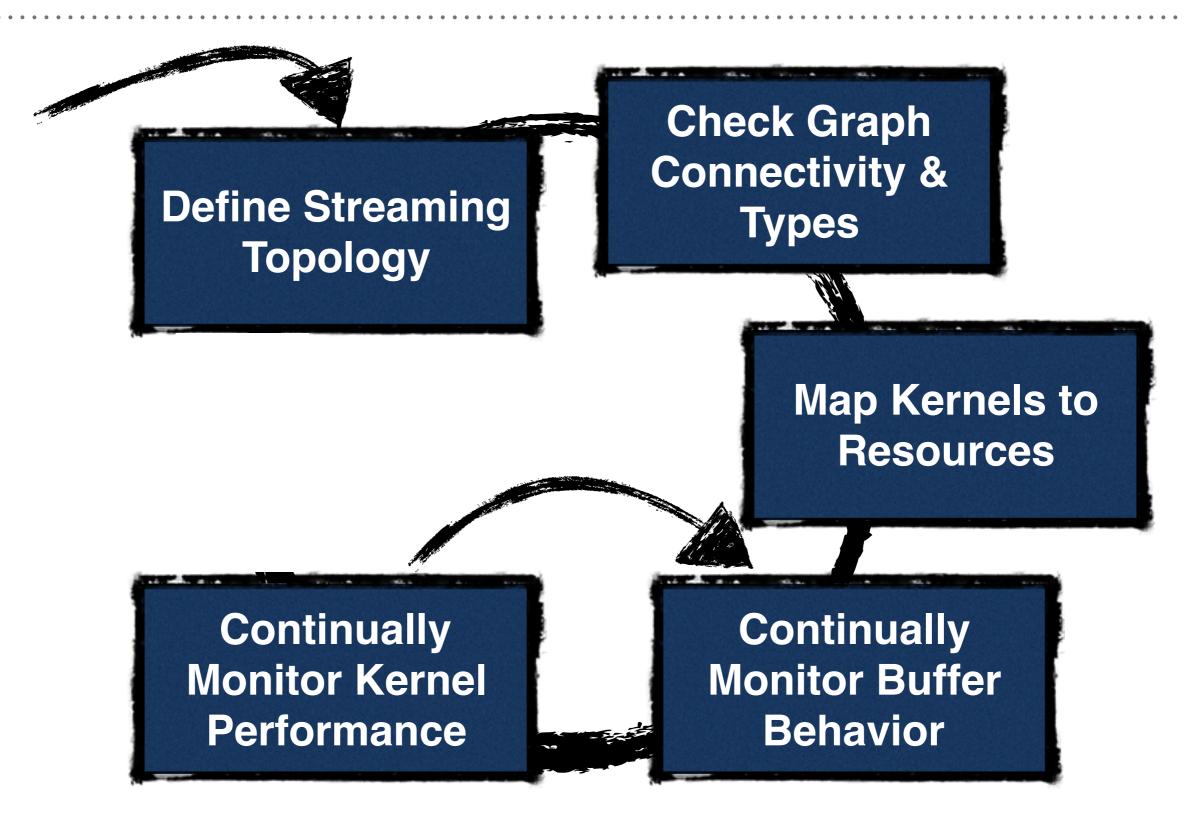
## STREAM PROCESSING (STRING SEARCH EXAMPLE)



### STREAM PROCESSING (STRING SEARCH EXAMPLE)



#### HOW IT WORKS (HIGH LEVEL)



```
template < class T > class search : public raft::kernel
ł
public:
   search( const std::string && term ) : raft::kernel(),
                                          term_length( term.length() ),
                                          term( term )
   {
      input.addPort< T >( "0" );
      output.addPort< std::size_t >( "0" );
   3
   search( const std::string &term ) : raft::kernel(),
                                        term_length( term.length() ),
                                        term( term )
   ł
      input.addPort< T >( "0" );
      output.addPort< std::size_t >( "0" );
   3
   virtual ~search() = default;
   virtual raft::kstatus run()
      auto &chunk( input[ "0" ].template peek< T >() );
      auto it( chunk.begin() );
      do
      ł
         it = std::search( it, chunk.end(),
                           term.begin(), term.end() );
         if( it != chunk.end() )
            output[ "0" ].push( it.location() );
            it += 1;
         }
         else
         Ł
            break;
         3
      3
      while( true );
      input[ "0" ].unpeek();
      input[ "0" ].recycle( );
      return( raft::proceed );
   }
private:
   const std::size_t term_length;
   const std::string term;
};
```

template < class T > class search : public raft::kernel

```
virtual raft::kstatus run()
{
    auto &chunk( input[ "0" ].template peek< T >() );
    auto it( chunk.begin() );
    do
    {
}
```

virtual ~search() = default;

```
it = std::search( it, chunk.end(),
                           term.begin(), term.end() );
         if( it != chunk.end() )
            output[ "0" ].push( it.location() );
            it += 1;
         }
         else
            break;
         3
      3
      while( true );
      input[ "0" ].unpeek();
      input[ "0" ].recycle( );
      return( raft::proceed );
   }
private:
   const std::size_t term_length;
   const std::string term;
```

```
};
```

ł

template < class T > class search : public raft::kernel

ł

```
search( const std::string && term ) : raft::kernel(),
                               term_length( term.length() ),
                               term( term )
 -{
    input.addPort< T >( "0" );
    outUDLLC
         search( const std::string && term ) : raft::kernel(),
  }
                                                          term_length( term.length() ),
  search
                                                          term( term )
 {
    inp
             input.addPort< T >( "0" );
    out
             output.addPort< std::size_t >( "0" );
  virtua
  virtua
         search( const std::string &term ) : raft::kernel(),
    aut
                                                       term_length( term.length() ),
    aut
    do
                                                       term( term )
    ł
         ł
             input.addPort< T >( "0" );
             output.addPort< std::size_t >( "0" );
         }
         break;
    while( true );
    input[ "0" ].unpeek();
    input[ "0" ].recycle( );
    return( raft::proceed );
private:
  const std::size_t term_length;
  const std::string term;
};
```

```
template < class T > class search : public raft::kernel
ł
public:
   search( const std::string && term ) : raft::kernel(),
                                         term_length( term.length() ),
                                         term( term )
   ł
      input.addPort< T >( "0" );
      output.addPort< std::size_t >( "0" );
   3
  search( const std::string &term ) : raft::kernel(),
                                       term_length( term.length() ),
                                       term( term )
   ł
      input.addPort< T >( "0" );
      output.addPort< std::size_t >( "0" );
   3
```

```
virtual ~search() = default;
```

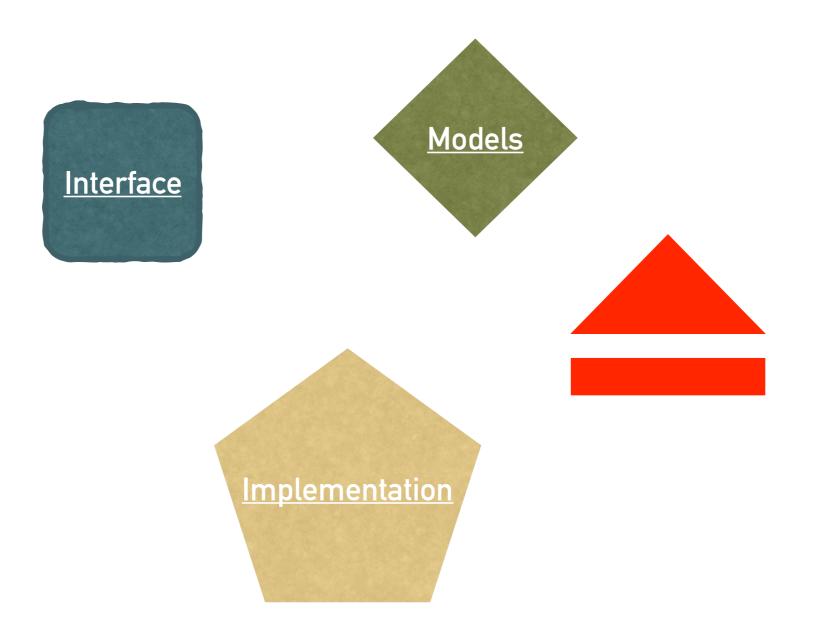
};

```
virtual raft::kstatus run()
   auto &chunk( input[ "0" ].template peek< T >() );
   auto it( chunk.begin() );
   do
   Ł
      it = std::search( it, chunk.end(),
                        term.begin(), term.end() );
      if( it != chunk.end() )
         output[ "0" ].push( it.location() );
         it += 1;
      }
      else
         break;
      3
   3
   while( true );
   input[ "0" ].unpeek();
   input[ "0" ].recycle( );
   return( raft::proceed );
const std::size_t term_length;
const std::string term;
```

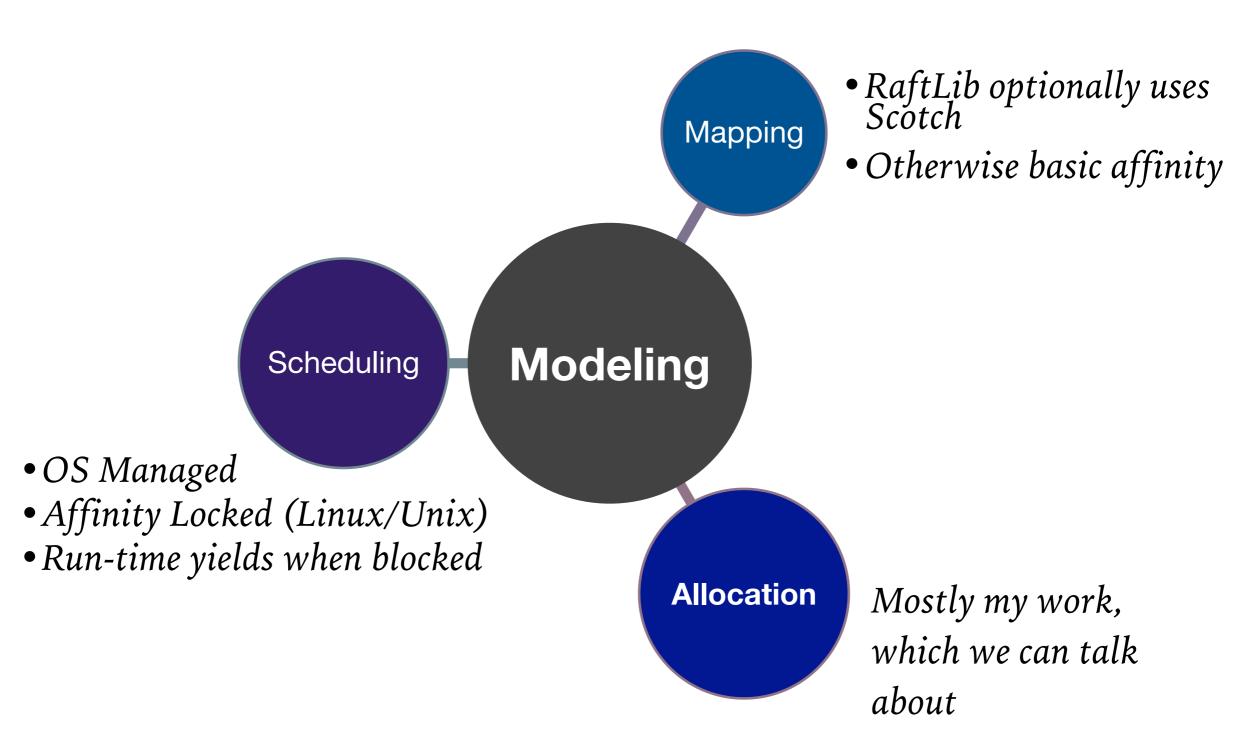
```
template < class T > class search : public raft::kernel
ł
public:
  search( const s virtual raft::kstatus run()
                  auto &chunk( input[ "0" ].template peek< T >() );
    input.addPor
    output.addPo
                  auto it( chunk.begin() );
                  do
  search( const s
                  ł
                      it = std::search( it, chunk.end(),
    input.addPor
    output.addPo
                                               term.begin(), term.end() );
                      if( it != chunk.end() )
  virtual ~search
  virtual raft::k
                          output[ "0" ].push( it.location() );
    auto &chunk(
                          it += 1;
    auto it( chu
    do
                      }
      it = std:
                      else
      if( it !=
        output
                          break;
        it +=
                      }
      else
                  }
        break;
                  while( true );
                  input[ "0" ].unpeek();
    while( true
    input[ "0" ]
                  input[ "0" ].recycle( );
    input[ "0" ]
    return( raft
                  return( raft::proceed );
  const std::size
  const std::string term;
};
```

```
int
main( int argc, char **argv )
ł
    using chunk = raft::filechunk< 256 >;
    using fr = raft::filereader< chunk, false >;
    using search = search< chunk >;
    using print = raft::print< std::size_t, '\n'>;
    const std::string term( "Alice" );
    raft::map m;
    fr read( argv[ 1 ], 1, term.length() );
    search find( term );
    print p;
    m += read >> raft::order::out >> find >> raft::order::out >> p;
    m.exe();
    return( EXIT_SUCCESS );
}
```

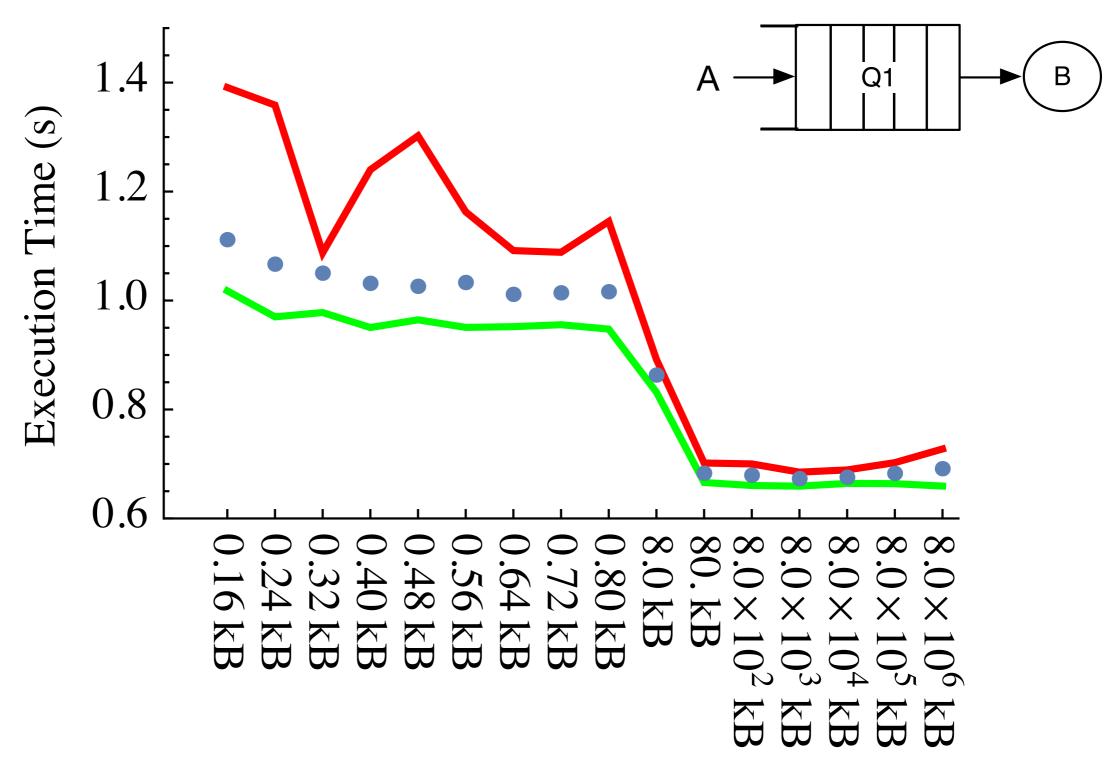
## **CHOOSE YOUR ADVENTURE**



#### MODELING



#### MODELING ISSUES – SIZE OF Q1

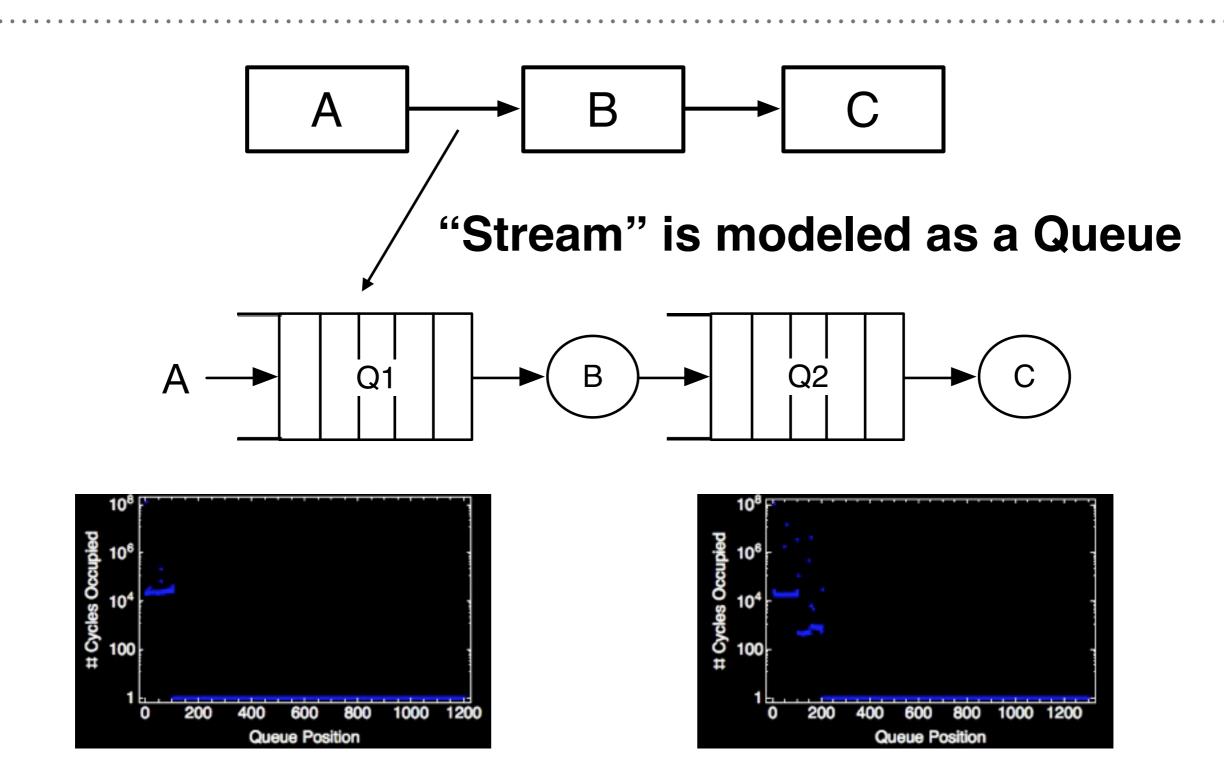


Buffer Size

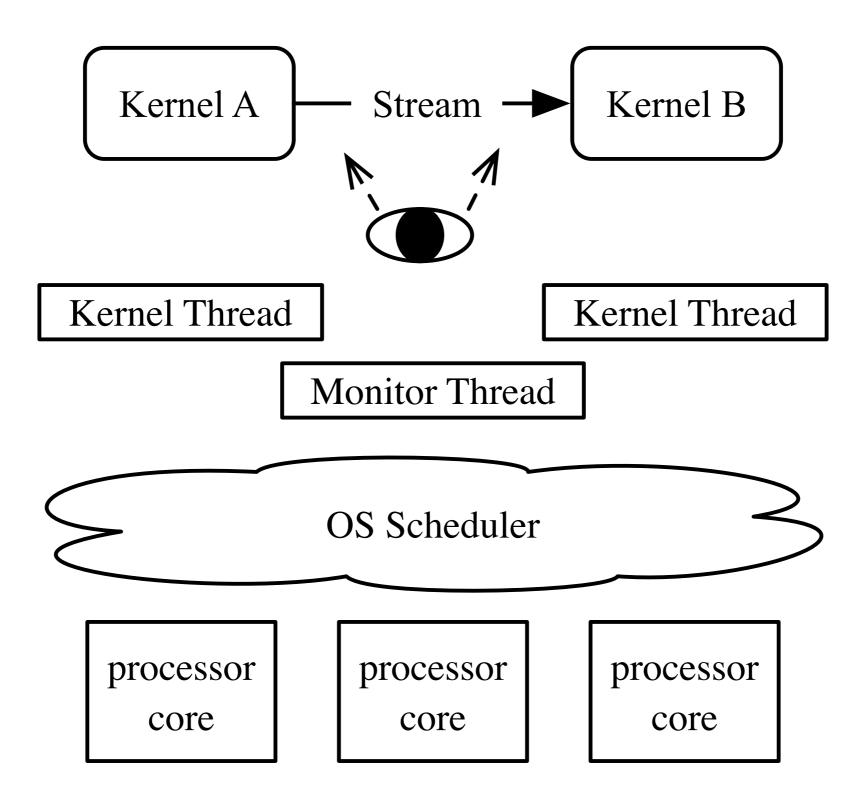
## WHY DO WE NEED TO SOLVE

- Buffer allocations take time and energy
- Programmers are horrible at deciding (too many parameters)
- ► Hardware specific locations matter (NUMA)
- Re-allocating with an accelerator takes even more time (bus latency, hand-shakes, etc.)
- Must be solved in conjunction with partitioning/scheduling/ placement

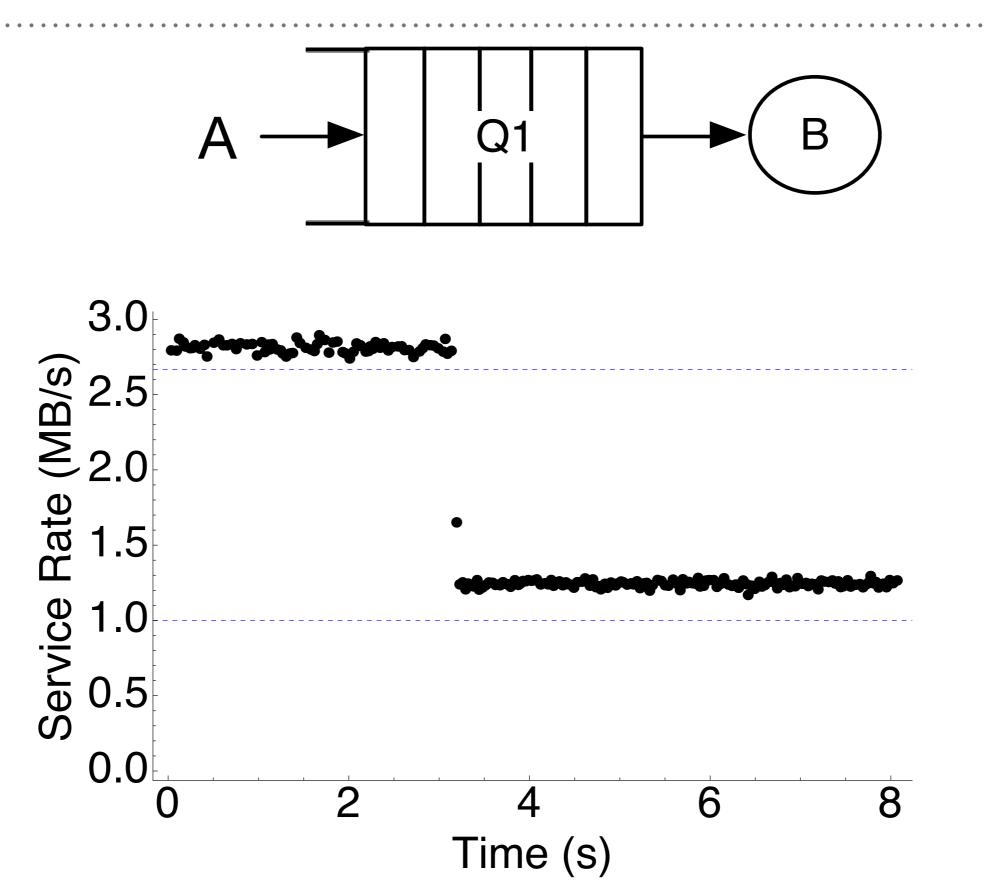
#### MODELING



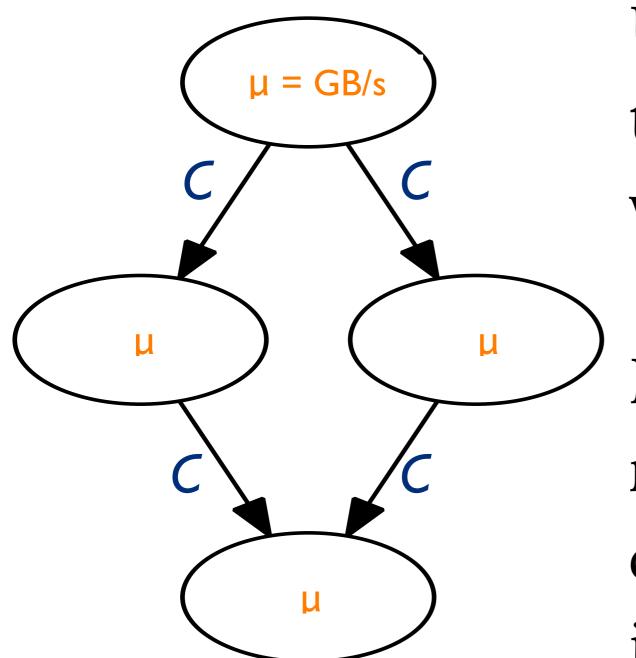
#### MONITOR



## **APPROXIMATE INSTRUMENTATION**



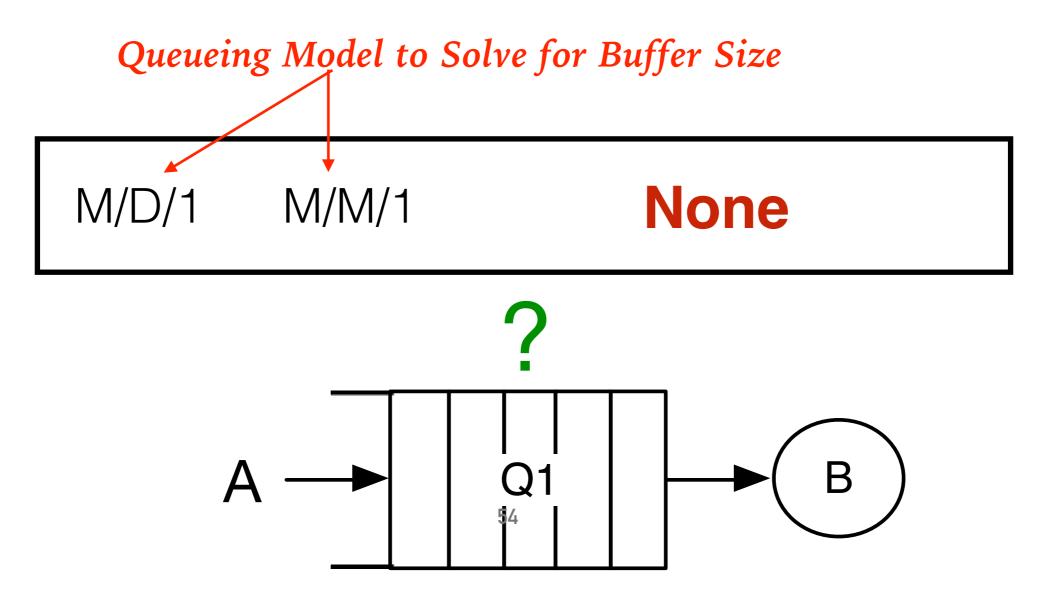
# SOLVE FOR THROUGHPUT QUICKLY



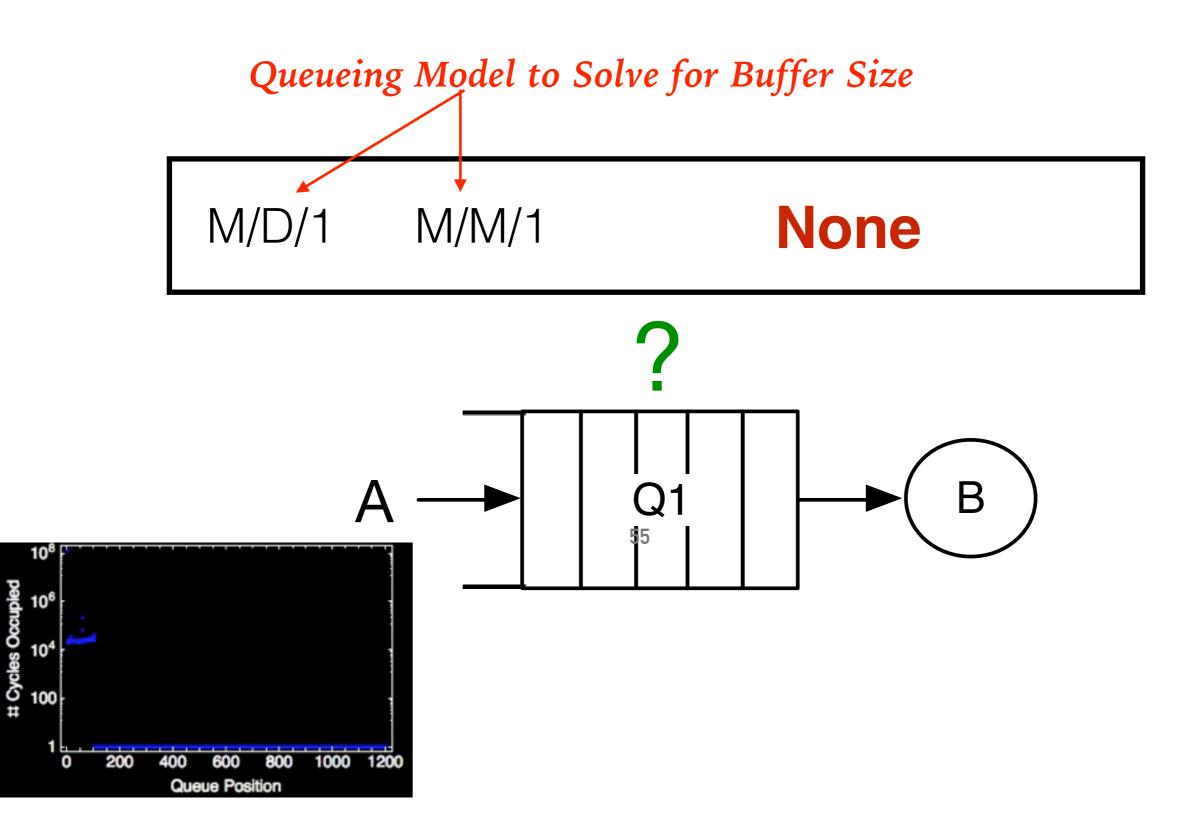
Use network flow model to quickly estimate flow within a streaming graph

Decompose queueing network and solve each queueing station independently

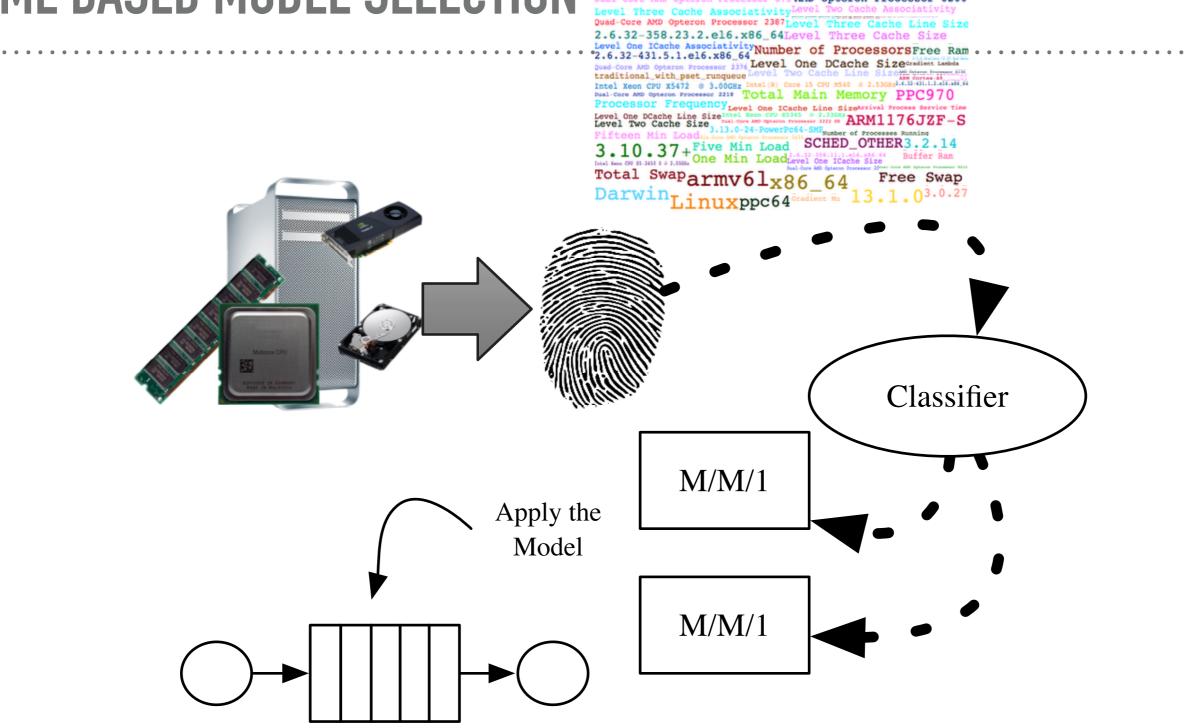
#### HOW WOULD YOU GET THE RIGHT BUFFER SIZE?



## HOW WOULD YOU GET THE RIGHT BUFFER SIZE?

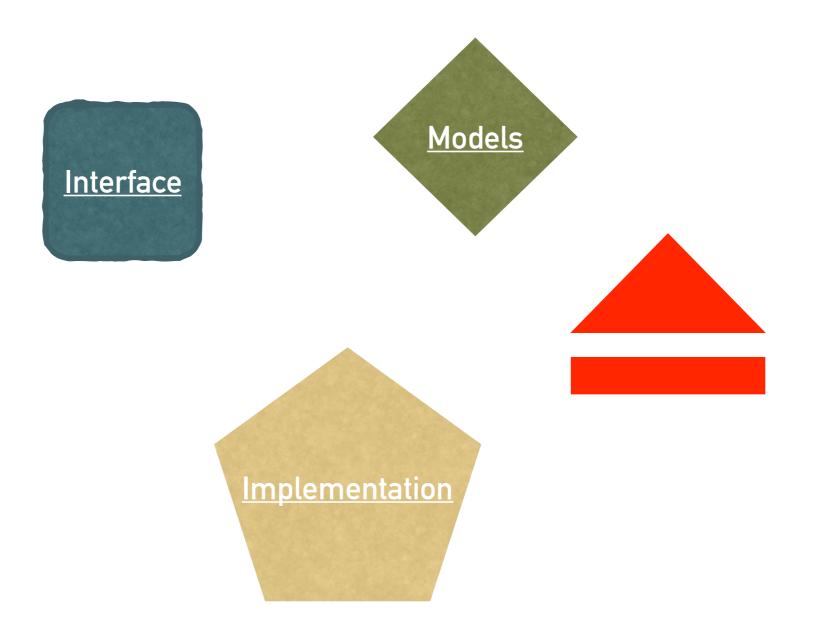


## **ML BASED MODEL SELECTION**

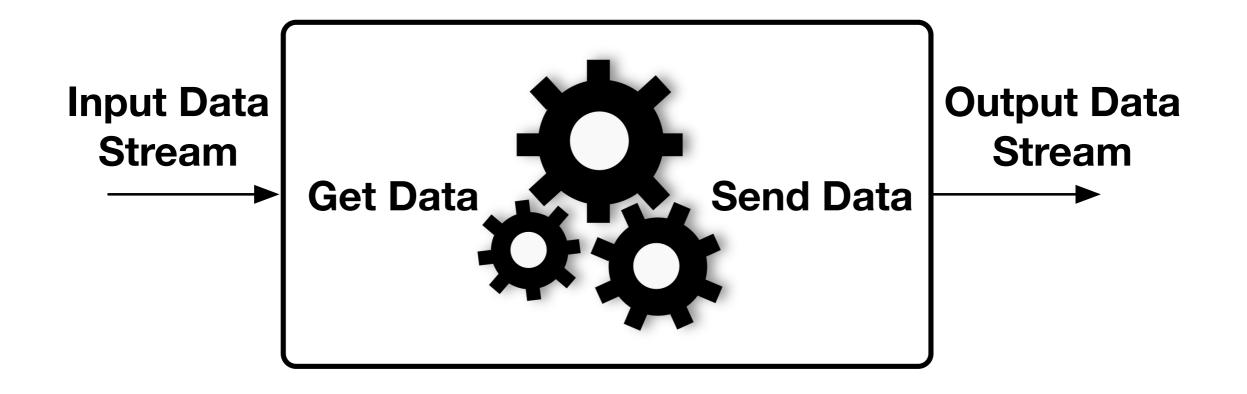


Dual Core AMD Opteron Processor 280Level One D Cache Associativity Dual Core AMD Opteron Processor 875 AMD Opteron Processor 6200

## **CHOOSE YOUR ADVENTURE**



## **COMPUTE KERNEL**



## **COMPUTE KERNEL**

```
class akernel : public raft::kernel
{
                                                                   Input Data
                                                                                     Output Data
public:
                                                                   Stream
                                                                                      Stream
                                                                                 end Data
    akernel() : raft::kernel()
    {
        //add input ports
        input.addPort< /** type **/ >( "x0", "x1", "x..." );
        //add output ports
        output.addPort< /** type **/ >( "y0", "y1", "y..." );
    }
    virtual raft::kstatus run()
    {
        /** get data from input ports **/
        auto &valFromX( input[ "x..." ].peek< /** type of "x..." **/ >() );
        /** do something with data **/
        const auto ret val( do something( valFromX ) );
        output[ "y..." ].push( ret val );
        input[ "x..." ].unpeek();
        input[ "x..." ].recycle();
        return( raft::proceed /** or stop **/ );
    }
};
```

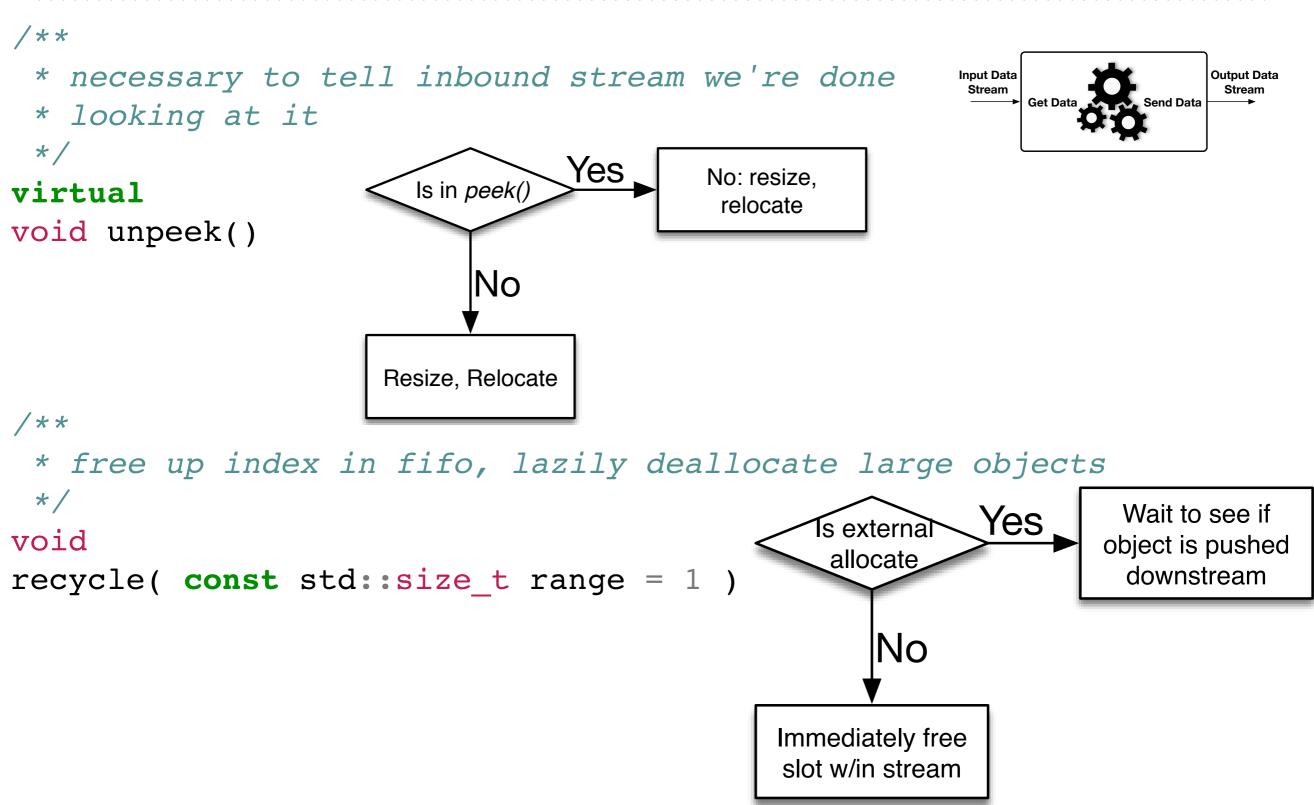
## **COMPUTE KERNEL**

```
class example : public raft::parallel k
                                                                Input Data
                                                                                  Output Data
                                                                 Stream
                                                                                   Stream
public:
                                                                     Get Dat
                                                                              end Data
   example() : parallel k()
   {
      input.addPort< /** some type **/ >( "0" );
      /** add a starter output port **/
      addPort();
   }
   /** implement virtual function **/
   virtual std::size t addPort()
   {
     return( (this)->addPortTo< /** type **/ >( output /** direction **/ ) );
   }
b
                 example
                                          b
                                                                C
```

## **RECEIVING DATA**

```
/**
 * return reference to memory on
                                                         Input Data
                                                                         Output Data
                                                          Stream
                                                                          Stream
 * in-bound stream
 */
template< class T >
T& peek( raft::signal *signal = nullptr )
template< class T >
autorelease< T, peekrange > peek range( const std::size t n )
• Returns object with "special access to stream"
• Operator [] overload returns auto pair
• Direct reference as in peek() for each element
                             template < class T > struct autopair
                             {
                                ጥ
                                                &ele;
                                Buffer::Signal &sig;
                             };
```

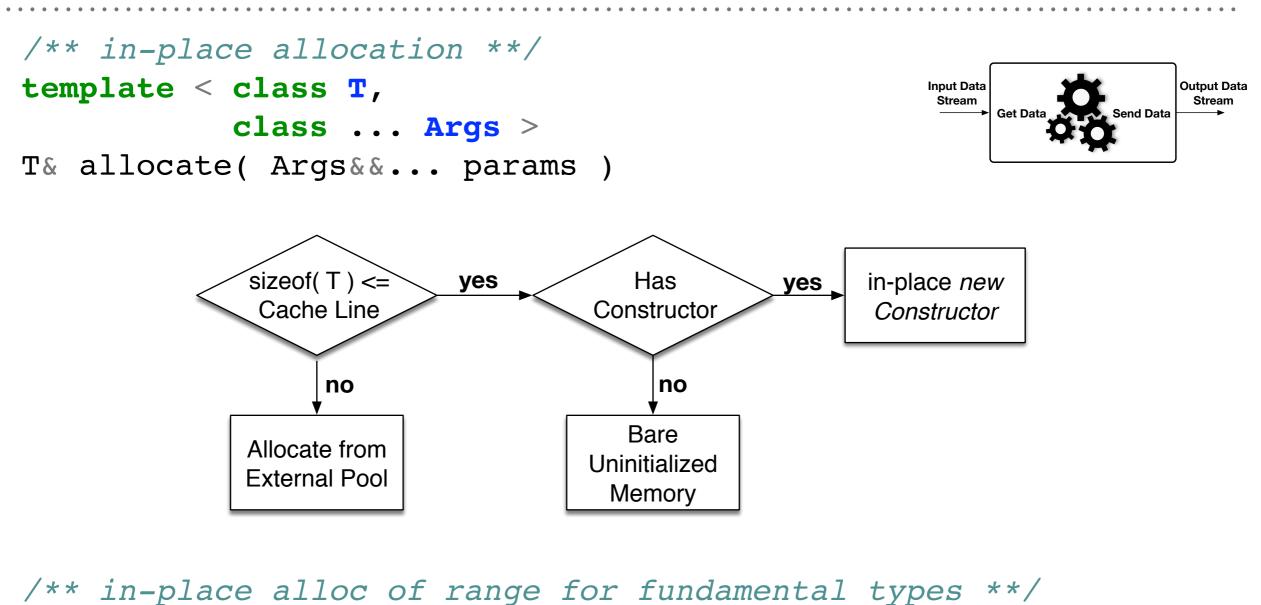
### **RECEIVING DATA**



## **RECEIVING DATA**

```
/**
 * these pops produce a copy
                                                  Input Data
                                                                 Output Data
                                                   Stream
                                                                 Stream
 */
template< class T >
void pop( T &item, raft::signal *signal = nullptr )
template< class T > using pop range t =
 std::vector< std::pair< T , raft::signal > >;
template< class T >
void pop_range( pop_range_t< T > &items,
                 const std::size t n items )
/**
 * no copy, slightly higher overhead, "smart object"
 * implements peek, unpeek, recycle
 */
template< class T >
autorelease< T, poptype > pop s()
```

#### **SENDING DATA**



```
template < class T >
auto allocate_range( const std::size_t n ) ->
std::vector< std::reference wrapper< T > >
```

## **SENDING DATA**

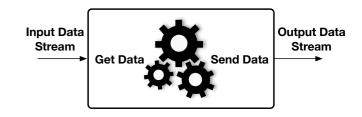
```
Input Data
Stream
Get Data
Get Data
Stream
```

```
/** release data to stream **/
virtual
void send( const raft::signal = raft::none )
/** release data to stream **/
virtual
```

```
void send_range( const raft::signal = raft::none )
```

```
/** oops, don't need this memory **/
virtual void deallocate()
```

#### **SENDING DATA**



/\*\* multiple forms \*\*/
template < class T >
void push( const T &item, const raft::signal signal = raft::none )

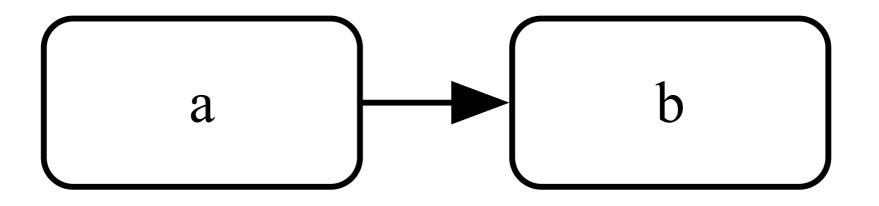
#### **INCLUDED KERNELS**

}

```
/**
 * thread safe print, specialization for '\n' vs. '\0'
 */
template < typename T, char delim = ' \setminus 0' > class print
/** read from iterator to streams **/
static
raft::readeach< T, Iterator >
read each( Iterator &&begin,
           Iterator & & end )
/** write from iterator to streams **/
template < class T, class BackInsert >
static
writeeach< T, BackInsert >
write each( BackInsert &&bi )
{
    return( writeeach< T, BackInsert >( bi ) );
```

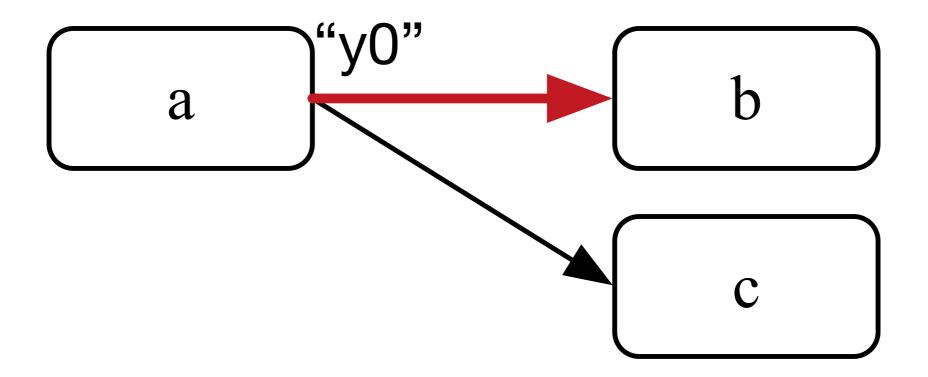
### **CONNECTING COMPUTE KERNELS**

raft::map m;
/\*\* example only \*\*/
raft::kernel a, b;
m += a >> b;

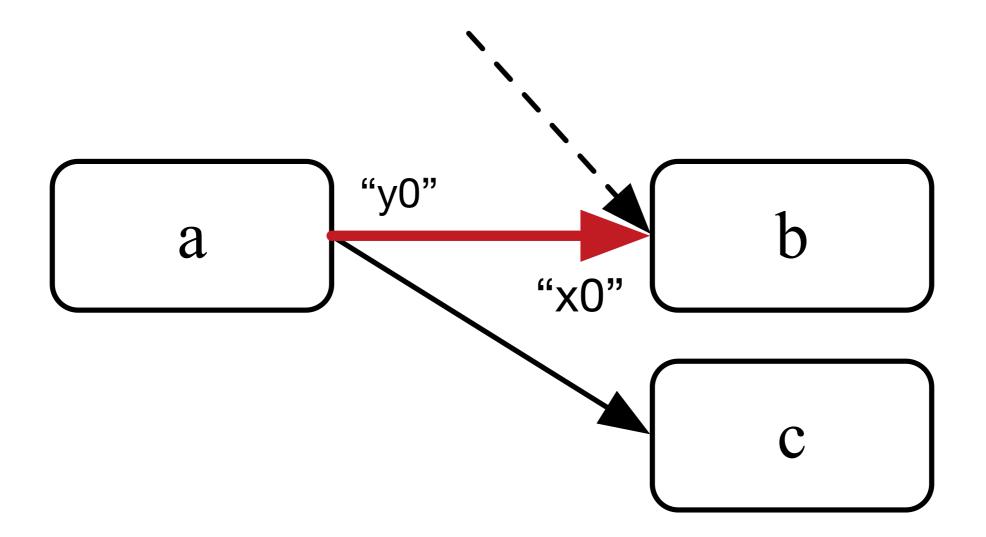


### **CONNECTING COMPUTE KERNELS**

raft::map m;
/\*\* example only \*\*/
raft::kernel a, b;
m += a[ "y0" ] >> b;



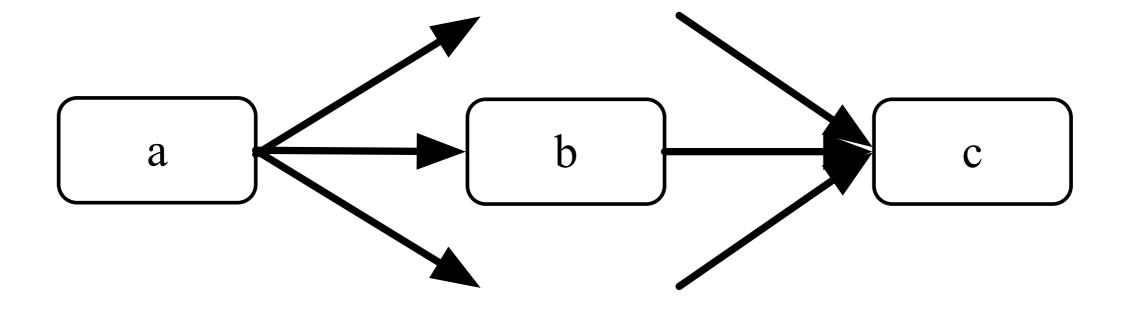
raft::map m;
/\*\* example only \*\*/
raft::kernel a, b;
m += a[ "y0" ] >> b[ "x0" ];



#### **CONNECTING COMPUTE KERNELS**

raft::map m;
/\*\* example only \*\*/
raft::kernel a, b, c;
m += a <= b >= c;

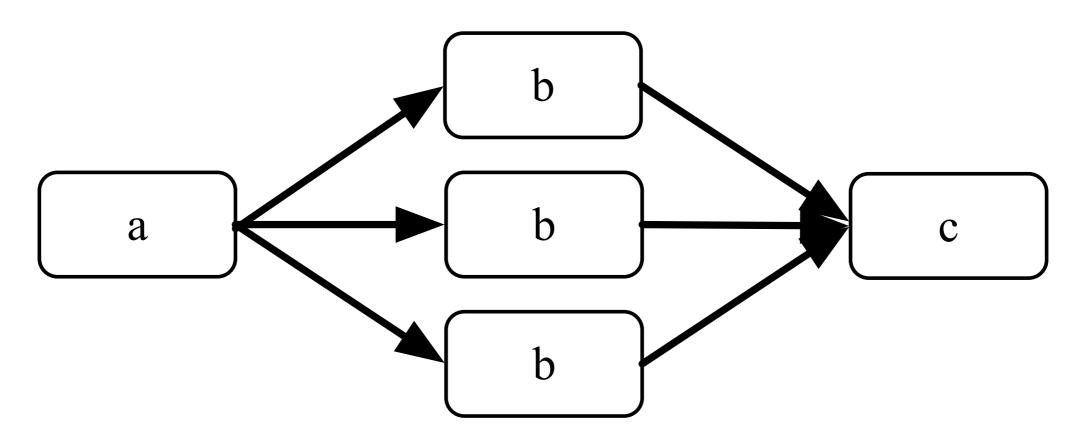
Topology user specifies



#### **CONNECTING COMPUTE KERNELS**

raft::map m;
/\*\* example only \*\*/
raft::kernel a, b, c;
m += a <= b >= c;

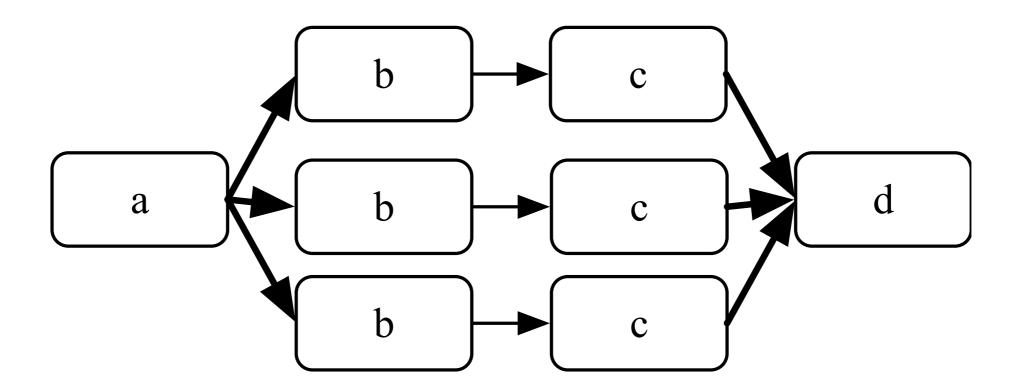
RaftLib Turns Into



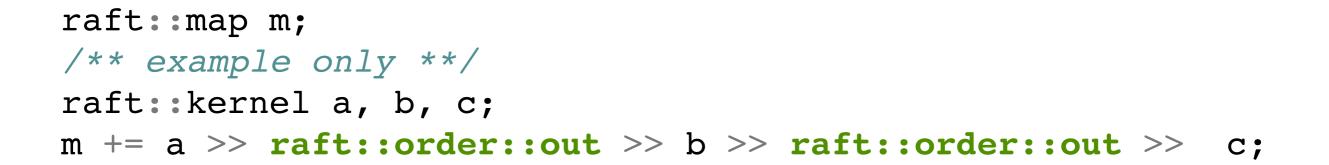
#### **CONNECTING COMPUTE KERNELS**

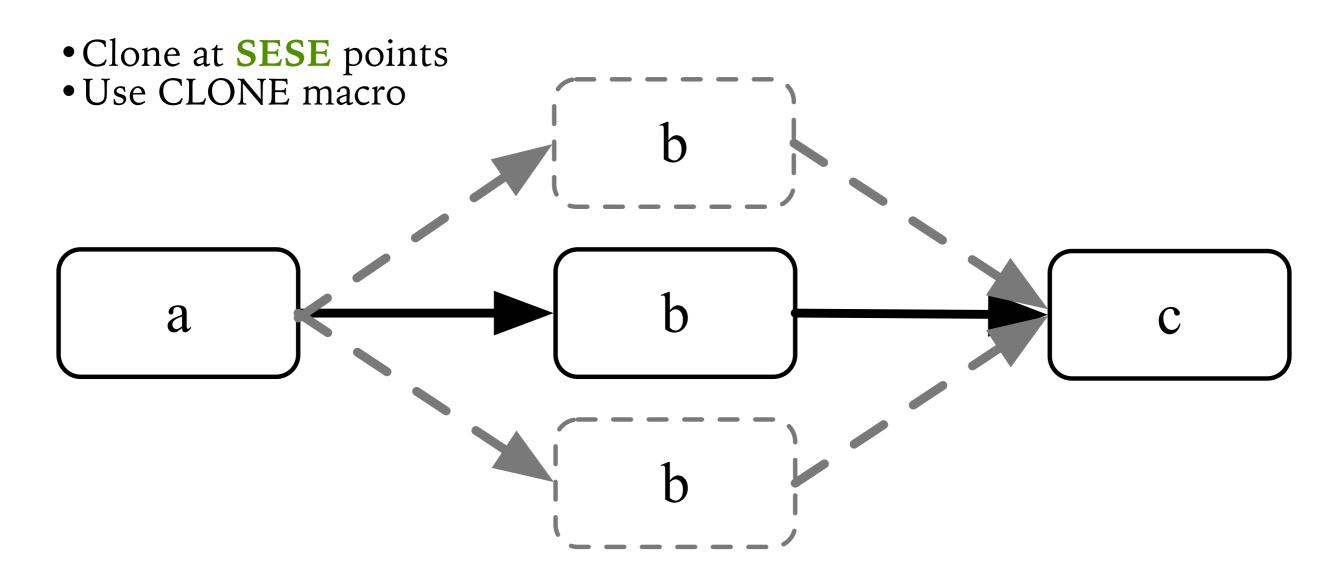
raft::map m;
/\*\* example only \*\*/
raft::kernel a, b, c, d;
m += a <= b >> c >= d;

RaftLib Turns Into



## **CONNECTING COMPUTE KERNELS**

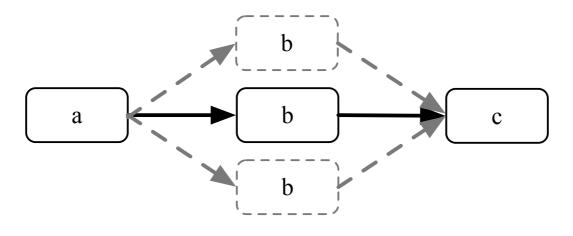




#### AUTO-PARALLELIZATION

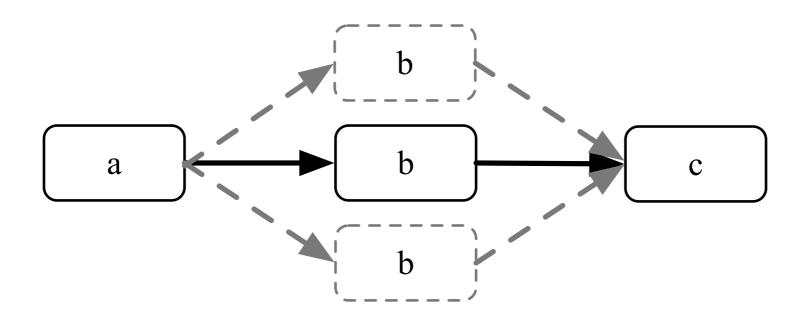
- 1) RaftLib figures out methodology
- 2) Runtime calls CLONE() macro of kernel "b"

```
#define CLONE()\
virtual raft::kernel* clone()\
{
    auto *ptr( \
        new typename std::remove_reference< decltype( *this ) >::type( ( *(\
        (typename std::decay< decltype( *this ) >::type * ) \
    this ) ) ));\
    /** RL needs to dealloc this one **/\
    ptr->internal_alloc = true;\
    return( ptr );\
}
```



## AUTO-PARALLELIZATION

- 3) Lock output port container of "a"
- 4) Register new port
- 5) Decide where to run it
- 6) Allocate memory for stream
- 7) Unlock output container of kernel "a"
- //do same on output side of "b"



# **CONNECTION TODO ITEMS**

► Better SESE implementation

Decide on syntax for set of kernels

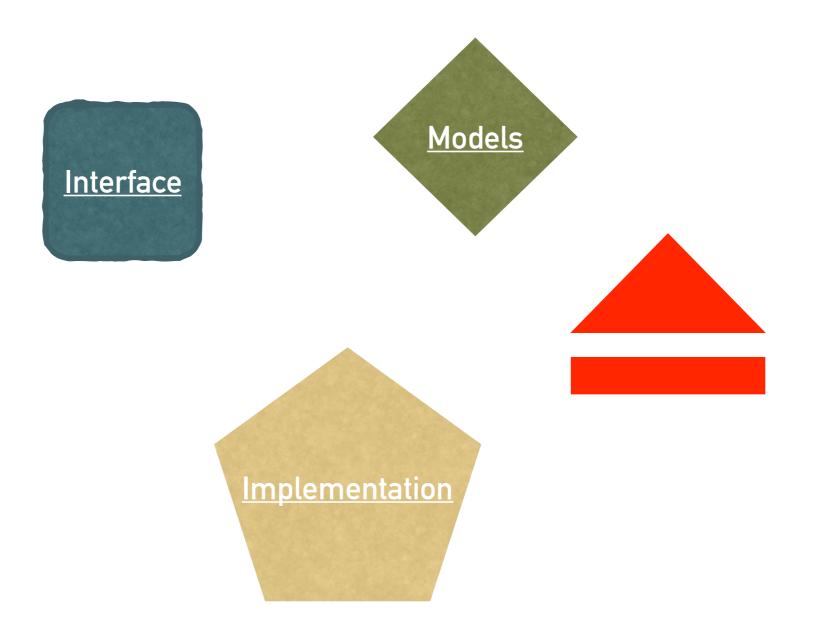
```
raft::map m;
/** example only **/
raft::kernel a, b, c, d;
m += src <= raft::kset( a, b, c, d ) >= dst;
```

Address space stream modifier (new VM space)

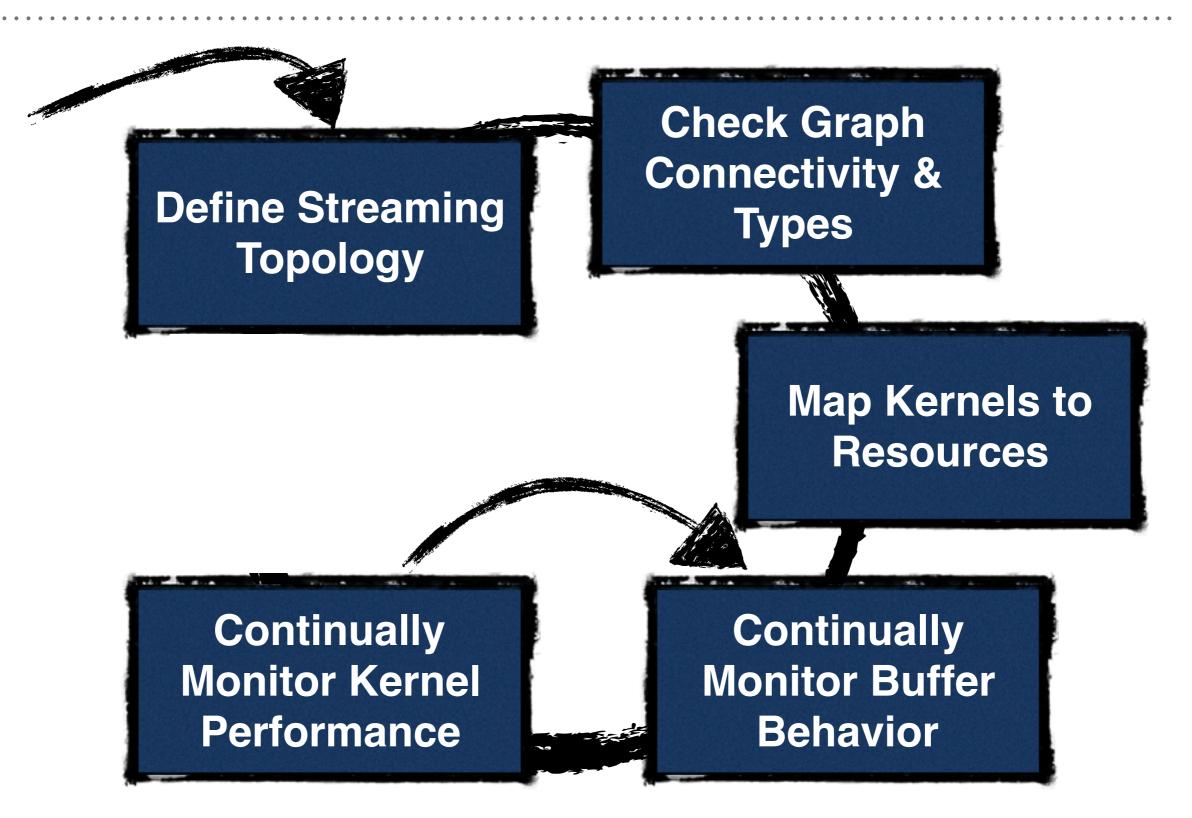
```
raft::map m;
/** example only **/
raft::kernel a, b, c;
m += a >> raft::vm::part >> b >> raft::vm::part >> c;
```

► Anything else?

### **CHOOSE YOUR ADVENTURE**

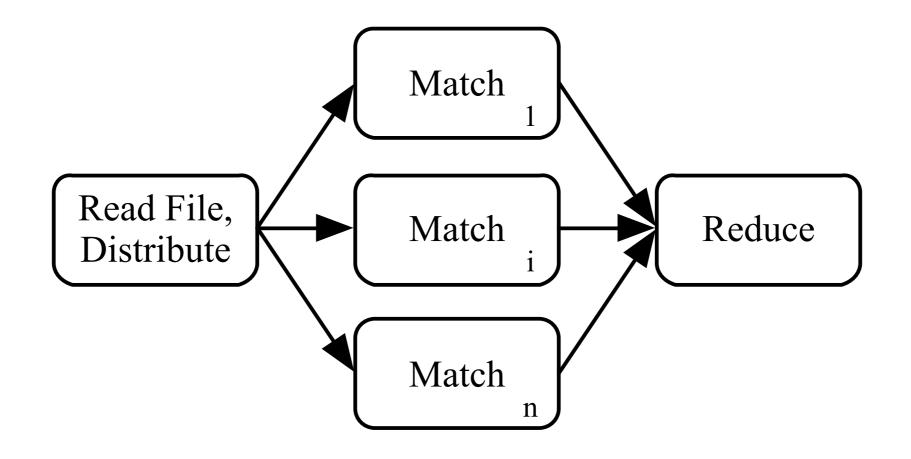


#### **IMPLEMENTATION DETAILS**



# **TOPOLOGY CHECK**

- ► Add kernels to map
- Check type of each link (potential for type optimization)
- Handle static split/joins (produce any new kernels)
- ► DFS to ensure no unconnected edges



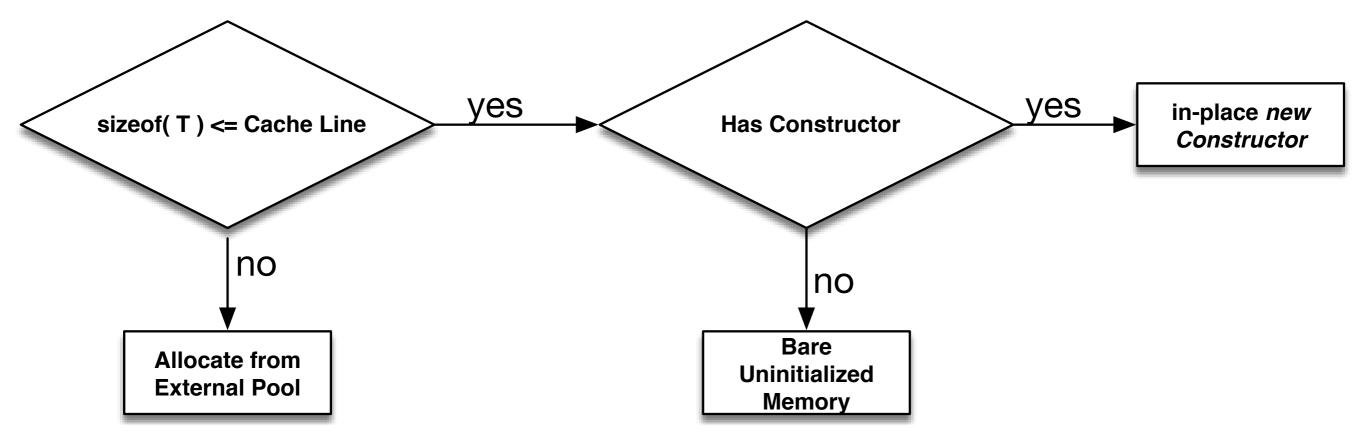
# PARTITION (LINUX / UNIX)

- Take RaftLib representation, convert to Scotch format
- Use fixed communications cost at each edge for initial partition
  - ► 2 main dimensions:
    - Flow between each edge in the application graph
    - Bandwidth available between compute resource
- Set affinity to partitioned compute cores
- ► Repartition as needed @ run-time
- TODO: incorporate OpenMPI utilities *hwloc* and *netloc* to get cross-platform hardware topology information

# **CHOOSE ALLOCATIONS**

#### ► Alignment

- SIMD ops often require memory alignment, RaftLib takes an alignment by default approach for in-stream allocations
- In-stream vs. External Pool Allocate (template allocators)



# MONITOR BEHAVIOR – ALLOCATORS

► Two options for figuring out optimal buffer size while running

- model based (discussed on modeling adventure path)
- branch & bound search
- separate thread, exits when app done
- ► pseudocode:

```
while( not done )
{
    if( queue_utilization > .5 )
    {
        queue->resize();
    }
    sleep( ALLOC_INTERVAL );
}
```

#### **RUN-TIME LOCK FREE FIFO RESIZING**



# **RUN-TIME LOCK FREE FIFO RESIZING**

Optimization...wait for the right conditions

```
if( rpt < wpt )
{
    //perfect to copy w/std::memcpy
}</pre>
```

#### ► Factory allocators

```
/** allocator factory map **/
std::map< Type::RingBufferType , instr_map_t* > const_map;
```

```
/** initialize some factories **/
const_map.insert( std::make_pair( Type::Heap , new instr_map_t() ) );
```

# MONITOR BEHAVIOR – PARALLELIZATION

Mechanics covered in interface, simple model here

Run in separate thread, term on exit

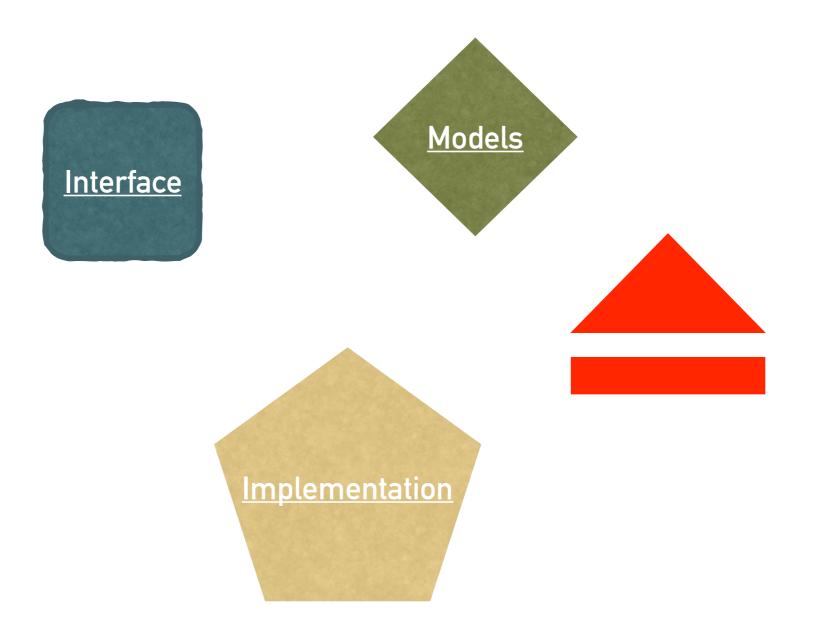
```
/** apply criteria **/
if( in_utilization > .5 && out_utilization < .5 )
{
    //tag kernel
    auto &tag( ag[ reinterpret_cast< std::uintptr_t >( kernel ) ] );
    tag += 1;
    if( tag == 3 )
    {
        dup_list.emplace_back( kernel );
    }
}
```

//after checking all kernels, handle duplication

# **IMPLEMENTATION TODO ITEMS**

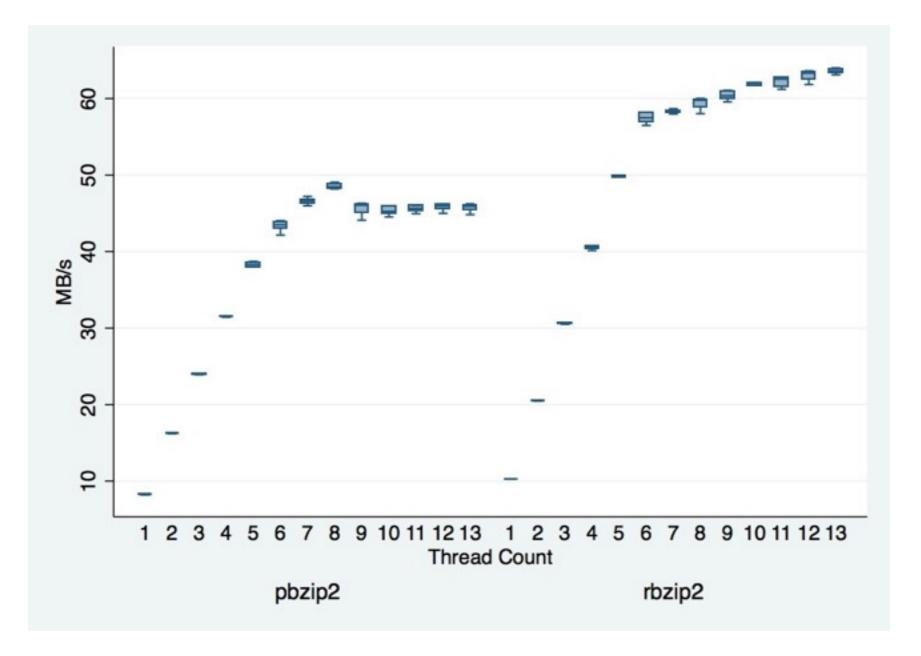
- Find fast SVM library to integrate (research code used LibSVM) for buffer model selection
- Integrate more production-capable network flow model for run-time re-partitioning choices
- Performant TCP links....
- ► RDMA on wish list
- QThreads Integration (see pool scheduler)
- *hwloc* and *netloc* integration (see partition\_scotch)
- Perf data caching (useful for initial partition)

### **CHOOSE YOUR ADVENTURE**



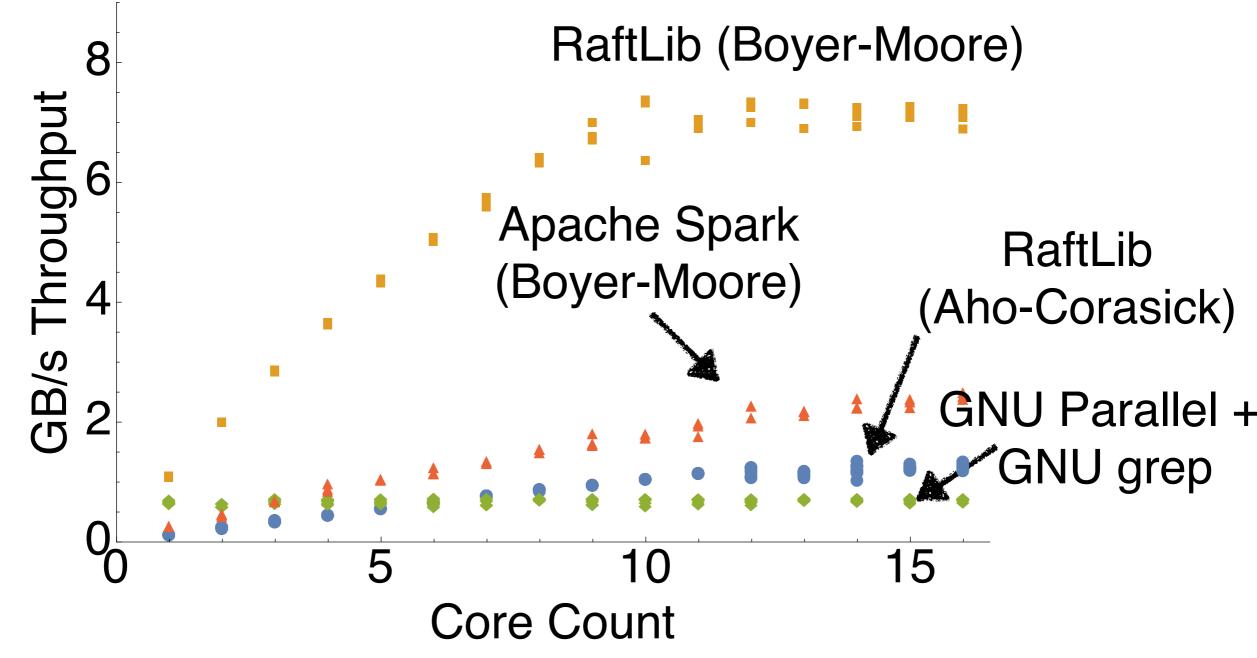
#### PERFORMANCE

- Decent compared to pthread stock implementation of pbzip2
- Parallel Bzip2 Example: <u>https://goo.gl/xyQAhm</u>



### PERFORMANCE

Fixed string search compared to Apache Spark, GNU Parallel + GNU Grep





#### **ABOUT ME**

my website

http://www.jonathanbeard.io



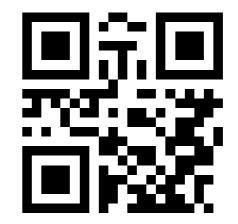
slides at

http://goo.gl/cwT5UB



project page

<u>raftlib.io</u>





#### video of talk given at #CppNow2016

http://goo.gl/mbxAwK



# **Stream Processing**

#### **Traditional Control Flow**

