

# Ct: C for Throughput Computing Channeling NeSL and SISAL through C++

# Mohan Rajagopalan Anwar Ghuloum

#### **Looking Backwards and Forwards**

- "There are no new ideas..."
  - Silicon trends introduce new opportunities to revisit
    - >Parallel programming models
    - >Parallel applications/algorithms
  - ...on a much different scale
- "...but much room for improvement..."
  - Modern programming methods require rethinking
    - >Dynamic compilation, managed runtimes
    - >Fine grained modularity
    - >Exceptionally complex and diverse patterns in single applications
      - Cf: Games!
- "...and new usages."
  - Parallel incremental/adaptive (re)computation
  - Forward scaling



# What Software Vendors are Telling Us

- Strong interest by ISVs for a parallel programming model which is:
  - Easy to use *and* high performance: sounds difficult already!
  - **Portable**: Desire the flexibility to target various HW platforms and adapt to future variations
- Programming parallel applications is 10,100,1000x\* <u>less</u> productive than sequential
  - Non-deterministic programming errors
  - Performance tuning is extremely microarchitecture-dependent
- Parallel HW is here today, better programming tools are needed to take advantage of these capabilities
  - Quad core on desktop arrived nearly a year months ago
  - Multi- and Many-core DP and MP machines are on the way
  - (Also, programmable GPUs going on 8 years)

\*Depends on which developer you ask.



#### Why We Started With Ct



- We moved from video algorithms to physics kernels
  - Rigid Body Dynamics
    - Broad and narrow-phase collision
    - Solvers
  - Cloth Simulation
- Found it painful to program using "legacy" parallel programming models
- Not surprisingly, same concerns as software vendors

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• (Nested) data parallel models make it easier

Ct: <6 lines of code, faster, scalable TVEC<I32> cols, TVEC<F64> v

TVEC<F64> expv, product, result; expv = distribute(v,cols); product = A\*expv; return addReduce(product);



#### **Irregular Data Structures**





4x4 sparse matrix

4 element vector of variable length vectors

Flattened representation with column & row metadata

- A classic example: Sparse matrices
  - Common in RMS applications
  - Difficult for a programmer to deal with

# Nested data parallelism handles irregular structures automatically



values

row

index

columns

#### Why Dataflow is Interesting

- Data isolation
  - Spatio-temporal localization of effects leads to desirable properties for parallelize
    - $\rightarrow$ Locality preserved
    - $\rightarrow$ Safety is guaranteed
- Required agility for many-core
  - Scaling
    - >Stretching "horizontally" to more threads, smaller footprints
    - >Stretching "vertically" to control memory bandwidth, arithmetic intensity
  - Adaptivity
    - >For incremental recomputation
    - >Intelligent, scalable synchronization/scheduling algorithms



# Language Vehicle for General Purpose Parallel Programming



Ct Api

- Nested Data Parallelism
- Deterministic Task Parallelism

Deterministic parallel programming
Fine grained concurrency and synch
Dynamic (JIT) compilation
High-performance memory management
Forward-scaling binaries for SSEx, ISAx

Parallel application library development

Performance tools for future architectures



# What Is Ct?

#### *"Extending"* C++ for Throughput-Oriented Computing

- Ct adds new data types (parallel vectors) & operators to C++
  - Library interface and is ANSI/ISO-compliant
- Ct abstracts away architectural details
  - Vector ISA width / Core count / Memory model / Cache sizes
- Ct forward-scales software written today
  - Ct platform-level API is designed to be *dynamically* retargetable to SSE, SSEx, ISA x, etc
- Ct is deterministic\*
  - No data races

Nested data parallelism and deterministic task parallelism differentiate Ct on parallelizing irregular data and algorithm



### The Ct Surface API: Nested Data Parallelism ++



#### **TVECs**

The basic type in Ct is a TVEC

- TVECs are managed by the Ct runtime
- TVECs are single-assignment vectors
- TVECs are (opaquely) flat, multidimensional, sparse, or nested
- TVEC values are created & manipulated exclusively through Ct API

```
Declared TVECs are simply references to immutable values
TVEC<F64> DoubleVec; // DoubleVec can refer to any vector of doubles
...
DoubleVec = Src1 + Src2;
...
DoubleVec = Src3 * Src4;
Assigning a value to DoubleVec doesn't modify the value
representing the result of the add, it simply refers to a new value.
```



#### **Ct In Action: C User Migration Path using Vector-style**







#### **Ct in Action: Kernel-style Programming with Ct Lambdas**





### The Ct Threading Model



#### **Dataflow is back!**

#### One way of looking at Ct:

#### A declarative way to specify complex task graphs

What we needed:

- Fine-grained concurrency and synchronization support
  - A bunch of lightweight tasks arranged in a dependency graph
- Novel optimizations and usage patterns
  - Reuse of task graph (called *future-graph*)
  - Incremental/adaptive update of FG

What we came up with:

- A super-lightweight *futures*-based threading abstraction
- Primitives for bulk creation of futures and complex synchronization
   → Building blocks for dataflow-style task graphs
- Composable first-class objects to enable dynamic optimization



#### **Feather-weight "Threads": Futures**

Futures: (Almost) stateless task



- API: Spawn & Read
- Futures can be in one of 3 states
  - Unevaluated: can be "stolen" or evaluated by reader
  - Evaluating: reader should wait for the result
  - Evaluated: reader can just grab the result
- Scheduled using distributed queues
  - Enqueued futures serviced by underlying worker threads
- Futures-creation about 2-3 orders of magnitude less expensive than thread creation



#### Simplifying Complexity through Data-parallel Patterns







#### **High-Level Primitives**

• Enable automatic dynamically configurable parallelism







#### **Future Graphs Reuse and Adaptivity**

- Abstraction for collectively manipulating about groups of futures
  - Generic reuse in code (esp. loops)
  - Play with funky scheduling algorithms
- 3 Basic operations: Creation, Instantiation, Evaluation



# **Task Parallelism in Ct**

Two options:

Futures and HSTs (Hierarchical, Synchronous Tasks)

- Futures
  - Basically, any Ct Function/Lambda can be spawned off as an parallel task (can include both scalar and vector code)
- HSTs
  - A sensible generalization of Bulk Synchronous Processes
  - Regions can be hierarchical
  - Bodies of tasks can be mix of data parallel and scalar code
- More details: offline

![](_page_18_Picture_10.jpeg)

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#### For more information: www.intel.com/go/Ct

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