Porting the CMS pixel reconstruction to Julia: preliminary results

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October 1st , 2024

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Project Context

- Evaluation of Julia as a language for High Energy Physics:
	- [General evaluations](https://doi.org/10.1007/s41781-023-00104-x)
	- [Jet clustering evaluations](https://github.com/JuliaHEP/JetReconstruction.jl)
- Patatrack pixel reconstruction:
	- Standalone application extracted from CMS software.
	- Pixel reconstruction: the process of identifying and reconstructing particle trajectories by analyzing data from pixel detectors.
	- Tested over the years on multiple CPU and GPU technologies (OpenMP, CUDA, HIP, SYCL, Kokkos, etc.).

Main Goal

To evaluate the feasibility of using Julia for large-scale HEP applications by:

- 1. Re-writing pixel track reconstruction algorithms, and comparing it with existing C++ application.
- 2. Test Julia's ability to handle multithreading, GPU acceleration, and overall flexibility in a realistic HEP environment.

Intermediate Steps

- 1. Port the Patatrack application to Julia.
- 2. Ensure its structure and output are similar to the C++ applications.
- 3. Monitor and optimize performance of the Julia implementation.

Interface between Modules and the Event

Porting C++ to Julia

- 1. Understand the execution of the Patatrack reconstruction software
- 2. C++ : #ifndef include("macro") module reconstruction
- 3. Zero Index, One Index
- 4. Type casting
- 5. Pointers No pointer
- 6. ES data reading
- 7. Reinterpret for 32 bit words
- 8. Performance tips

Module Inclusion and Management in Julia

No Preprocessor: Julia doesn't use #include or #ifdef

directives like C++

File Inclusion: The include ("file.jl") function reads and executes files at runtime. Including the same file multiple times

can cause errors.

Managing Modules in Patatrack:

Single Point of Inclusion: Each module is included once in the Patatrack package to prevent multiple inclusions.

Controlled Access: Other parts of the program access these modules through Patatrack, ensuring no duplicate file inclusions.

Zero Indexing to One Indexing

All detector metadata (e.g., module indices) are zeroindexed in C++, but Julia uses 1-based indexing, requiring careful adjustment by incrementing indices by one during the transition.

```
nodule gpuClusterCharge
   include("../CUDACore/cuda_assert.jl")
   # using .gpuConfig
   include("../CUDACore/prefix scan.jl")
   using .prefix_scan:block_prefix_scan
   include("../CUDADataFormats/gpu clustering constants.jl")
   using . CUDADataFormatsSiPixelClusterInterfaceGPUClusteringConstants. pixelGPUConstants: INV ID, MAX NUM CLUSTERS PER MODULES, MAX NUM MODULES
   using Printf
   function cluster_charge_cut(id, adc, moduleStart, nClustersInModule, moduleId, clusterId, numElements)
       charge = fill(0, MAX NUM CLUSTERS PER MODULES) # mok = fill(0, MAX NUM CLUSTERS PER MODULES) # mnewclusId = fill(0, MAX_NUM_CLUSTERS_PER_MODULES) # m
       firstModule = 1endModule = moduleStart[1]for \mod \in firstModule:endModulefirstFixed = moduleStart[1 + mod]this ModuleId = id[first Pixel]
           @assert thisModuleId < MAX NUM MODULES
           @assert thisModuleId == moduleId[mod]
           nClus = nClustersInModule[thisModuleId+1]if nClus == 0if nClus > MAX_NUM_CLUSTERS_PER_MODULES
               @printf("Warning too many clusters in module %d in block %d: %d > %d\n",
             thisModuleId,
             Θ,
              nClus,
             MaxNumClustersPerModules)
           first = firstPixel
```
Module Error Toy Example

ERROR: MethodError: no method matching TurnOn(::Ferrari)

Closest candidates are: TurnOn(::Main.CarFunctions.CarFactory.Ferrari)

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Avoiding Pointers

// check CRC bit

const $uint64_t^*$ trailer = reinterpret_cast<const $uint64_t^*)$ (rawData.data()) + (nWords - 1); if (not errorcheck.checkCRC(errorsInEvent, fedId, trailer, errors_)) { continue;

// check headers

const $uint64_t^*$ header = reinterpret_cast<const $uint64_t^*$ >(rawData.data()); $header -$; bool moreHeaders = $true$; while (moreHeaders) { header++; bool headerStatus = errorcheck.checkHeader(errorsInEvent, fedId, header, errors_); $moreHeaders = headerStatus;$

// check trailers

bool more $\text{Trailers} = \text{true}$; trailer++; while (moreTrailers) { trailer--; bool trailerStatus = errorcheck.checkTrailer(errorsInEvent, fedId, nWords, trailer, errors_); $moreTrailers = trailerStats;$

const $uint32_t^*$ bw = (const $uint32_t^*)$ (header + 1); const $uint32_t^*$ ew = (const $uint32_t^*)$ (trailer);

$assert(\theta == (ew - bw) % 2);$

wordFedAppender_->initializeWordFed(fedId, wordCounterGPU, bw, (ew - bw));

```
wandCountandDBL = (aw - bu)
```
Reinterpret Function

C++20 std::bit_cast : https://en.cppreference.com/w/cpp/numeric/bit_cast

Performance

Initial Results

1000 Events / 95.2 Seconds = 10.5 Events / Second

Initial Results - Cont'd

1000 Events / 95.2 Seconds = 10.5 Events / Second only on Clusterizer

Heap allocations reducing time performance !!

Julia Documentation

523.0007221951678

```
julia> @time sum_global()
 0.011539 seconds (9.08 k allocations: 373.386 KiB, 98.69% compilation time)
523.0007221951678
julia> @time sum_global()
 0.000091 seconds (3.49 k allocations: 70.156 KiB)
```
"Unexpected memory allocation is almost always a sign of some problem with your code, usually a problem with type-stability or creating many small temporary arrays. Consequently, in addition to the allocation itself, it's very likely that the code generated for your function is far from optimal. Take such indications seriously and follow the advice below."

Vector to SVector


```
moreHeaders = truewhile moreHeaders
   headerStatus =check header (error check, errors in event, fed id, header
   moreHeaders = headerStatus
    if moreHeaders
       header byte start += 8header = (data(rawData))[header byte start: header byte start+7])
moreTrainer = truewhile (moreTrailer)
   trailerStatus = check_trailer(error_check,errors_in_event, fed_id, n_wor
   moreTrailer = trailerStatus
    if moreTrailer
       trailer byte start -8trailer = (dataFormats.data(rawData),trailer_byte_start:trailer_byt
```
Before

Before

Views For Slices

In Julia, an array "slice" expression like array[1:5, :] creates a copy of that data (except on the left-hand side of an assignment, where $array[1:5, :] = ...$ assigns in-place to that portion of array).

```
moreHeaders = true
  ile moreHeaders
   headerStatus = check header(error check, errors in event, fed id, header,
   moreHeaders = headerStatus
   if moreHeaders
       header byte start += 8header = \omegaviews (data(rawData)[header byte start:header byte start+7]
                                                                                 After
moreTrain = truehile (moreTrailer)
   trailerStatus = check_trailer(error_check,errors_in_event, fed_id, n_words
   moreTrailer = trailerStatus
   if moreTrailer
       trailer byte start -8trailer = @views (dataFormats.data(rawData),trailer_byte_start:trailer
                                                                                After
```
Abstract Type Within Struct

Type Instability: Type of member Variable not known at compile time, Compiler Allocates Extra memory on Heap. Dynamic Dispatch due to runtime type check slows down performance

After Other Optimizations…

1828 decoding 58026 digis. Max is 300000 1807 Processed 1000 events in 1.461950e+01 seconds, throughput 68.4018 events/s, CPU usage per thread: 68.5% <haled47@khaled47-virtual-machine:~/Desktop/cernProject/pixeltrack-standalone\$

Run Time Drops to 19.7 seconds (9 seconds on Ayman's Macbook M1 Processor) Run Time C++ 14.6 seconds (6 seconds on Ayman's Macbook M1 Processor)

The Process

Achievements

Patatrack 16th Hackathon Results

Patatrack 16th Hackathon Results- Cont'd

Clusterizer Validation

Cluster Ids of digis Validated

RecHits Validation

[mcharaf@lxplus928 PROJECT]\$ diff RecHitsJulia.txt RecHitsC++.txt [mcharaf@lxplus928 PROJECT]\$

Doublets Validation

[mcharaf@lxplus928 PROJECT]\$ diff DoubletsJulia.txt DoubletsC++.txt [mcharaf@lxplus928 PROJECT]\$

Running Time: Up to Doublets

C_{++} :

Processed 1000 events in 3.180145e+01 seconds 31.8 Seconds

Julia:

35.081 s (235105557 allocations: 76.85 GiB) Terminal will be reused by tasks, press any key to close it.

35.1 Seconds

Ported 100% of the local reconstruction

Ongoing Activities:

• **Precompilation and Distribution**

▪ **PackageCompiler.jl**

• **Multithreading**

- **Threads.@threads**
- **FLoops.jl**
- **ThreadPools.jl**
- **GPU Integration**
	- **CUDA.jl**

Next Steps and Future Work

- Integrate GPU acceleration into the entire application.
- Optimize Performance.
- Share findings to encourage more adoption of Julia in scientific computing!

Conclusion

• **Advantages gained from using Julia**

- o Syntax that is easier to read and write
- Automatic memory management
- o Large Ecosystem

• **Key Accomplishments**

o Significant strides in initial implementations and testing.

• **Remaining Challenges**

Address multithreading, pre-compilation, and GPU integration.

• **Future Vision**

o Enhance both development speed and runtime performance.

Thank You!

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