Accelerating Java Programs on RISC-V with Vector Instructions via TornadoVM and OCK

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The University of Manchester

J Extension Task Group
27th February 2025

TORNADOVM

Outline

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1. Motivation
2. Overview of TornadoVM
3. Java Acceleration on RISC-V
   - TornadoVM and oneAPI Construction Kit (OCK)
4. OCK for RISC-V
5. Performance Numbers on RISC-V RVV 1.0 – Feb 2025
6. Conclusions & Discussions

Who am I?

Dr. Juan Fumero

Research Fellow @ University of Manchester

Architect and Developer of TornadoVM

oneAPI **Intel Innovator**

- oneAPI Lang SIG
- oneAPI Hardware SIG

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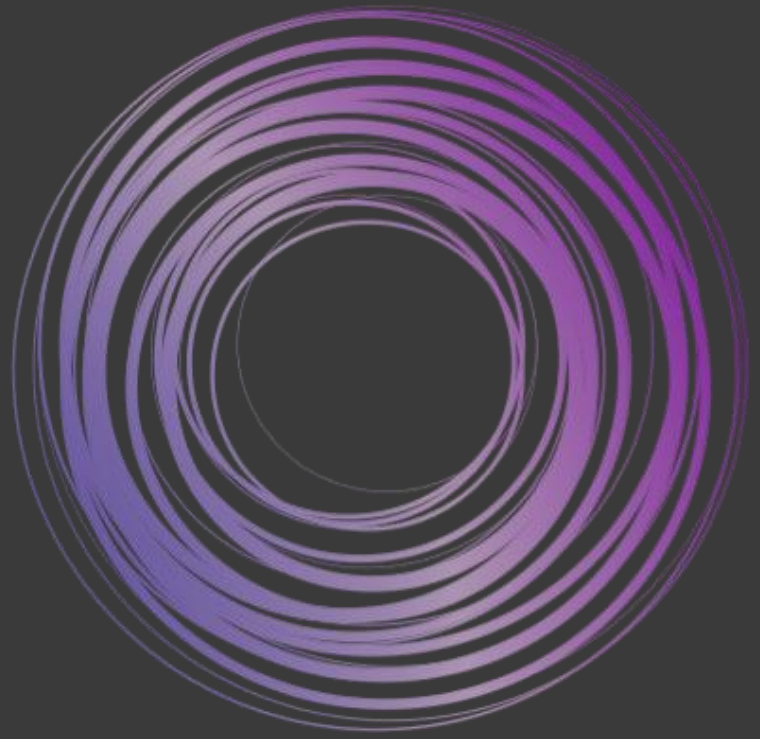
Background:

PhD: Java JIT Compilers for GPUs

Oracle Labs *GraalVM/Truffle*

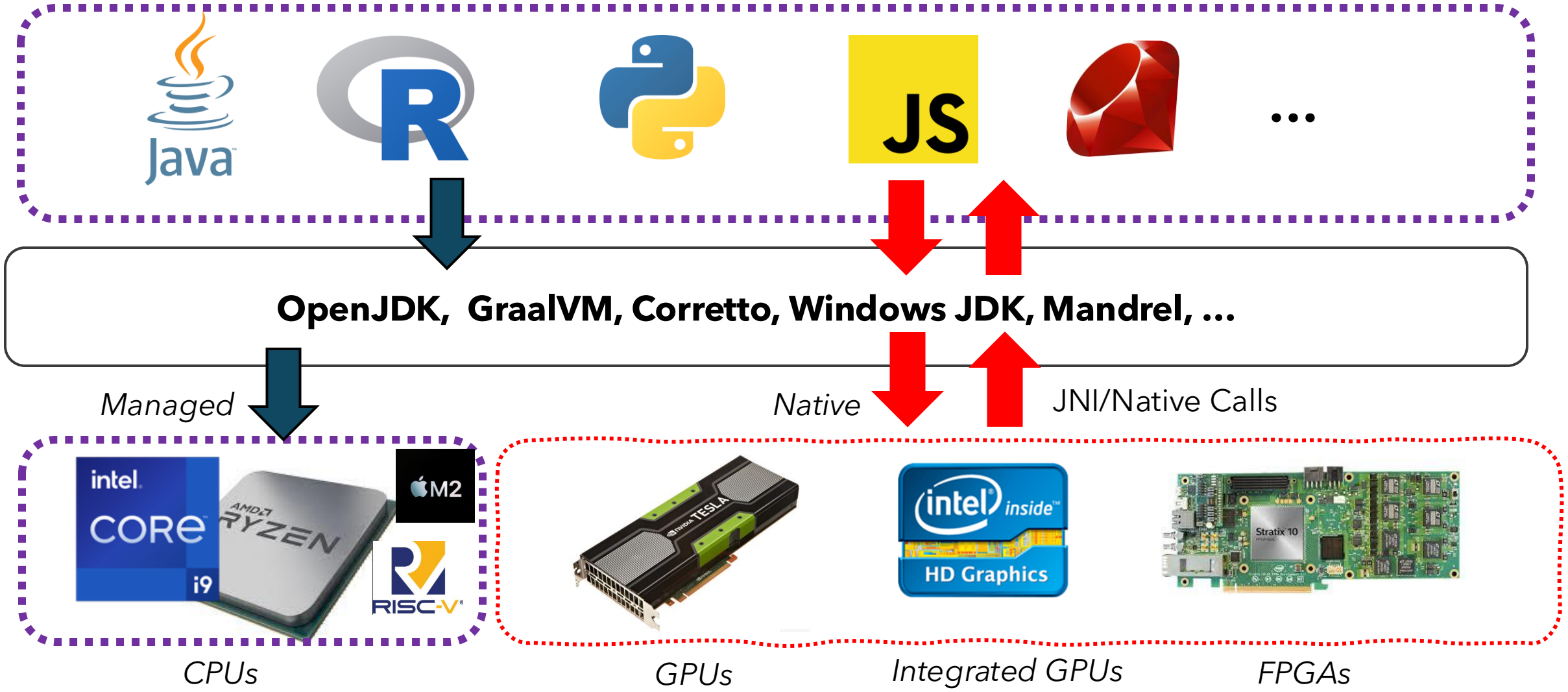
Intel CilkPlus Vectorization for Root and GeantV



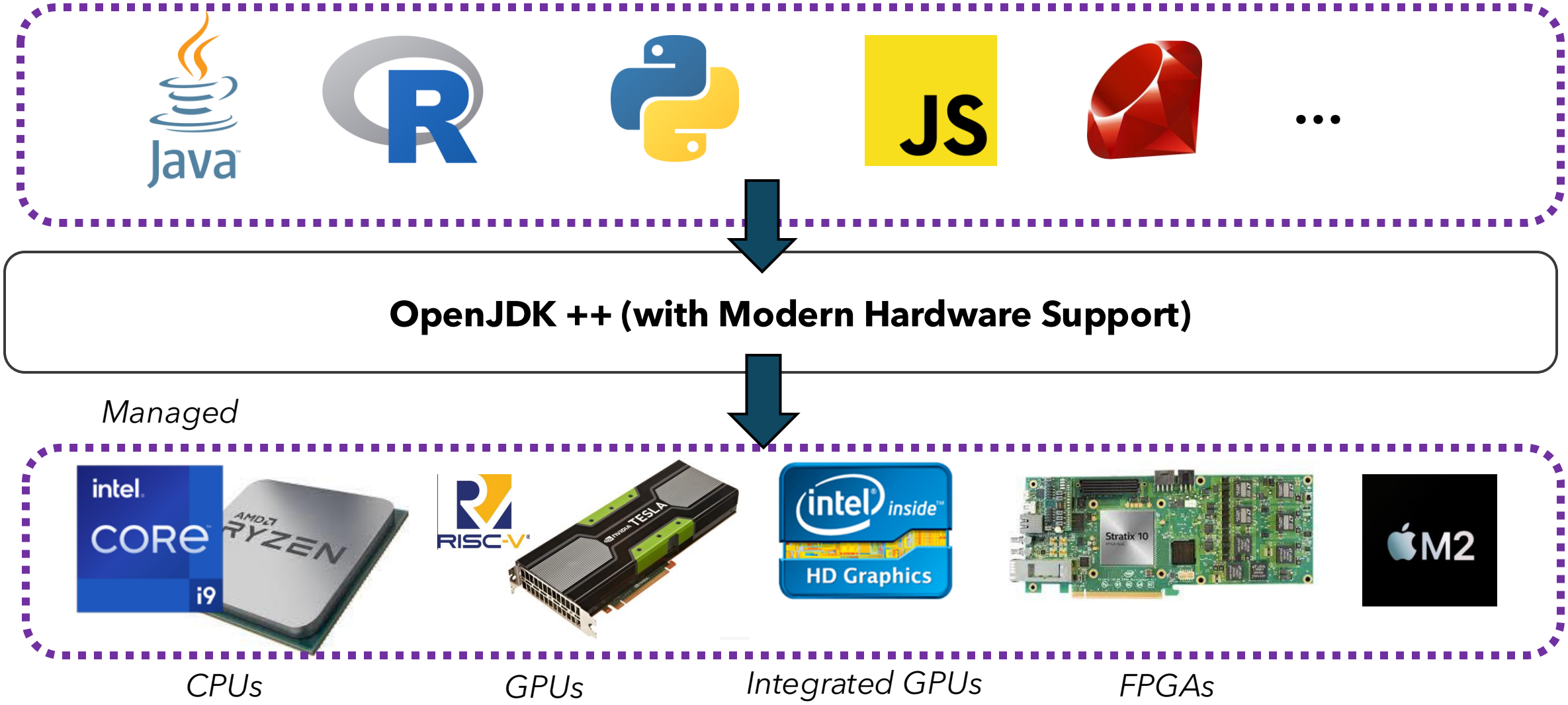
www.tornadovm.org

Motivation

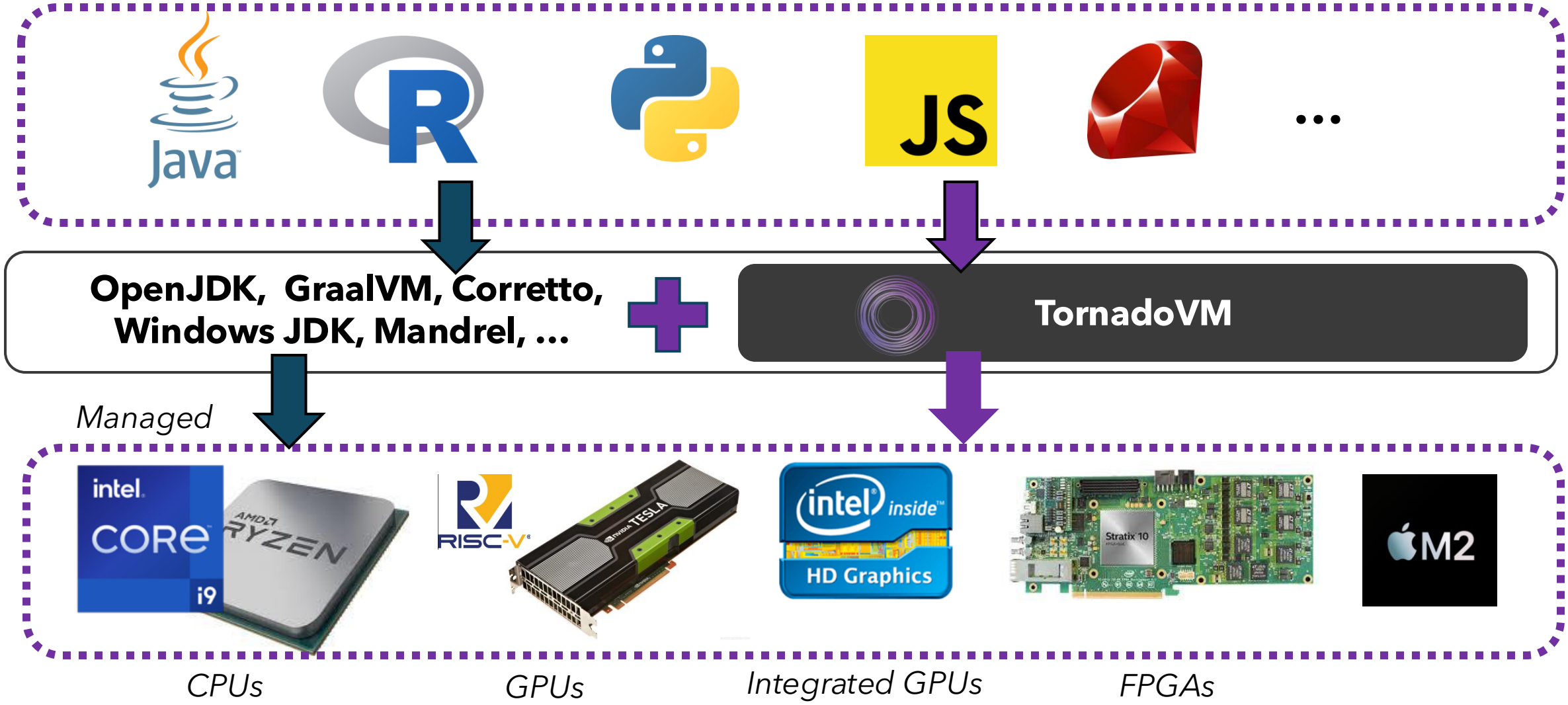
Enabling Acceleration for Managed Runtime Languages



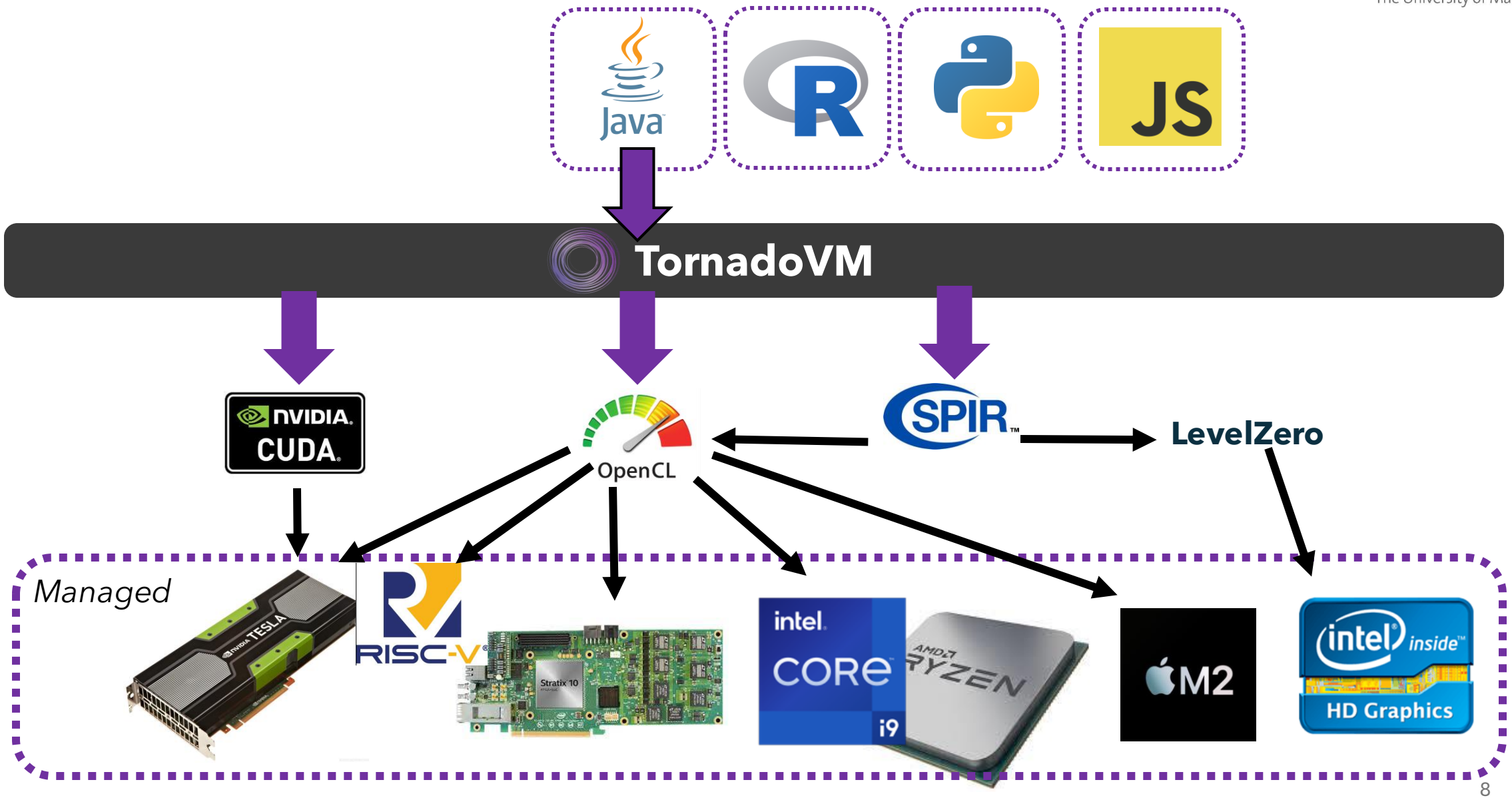
But what if?


Java
R
JS
...
OpenJDK ++ (with Modern Hardware Support)
Managed
intel
CORE
i9
AMD
RYZEN
RISC-V
TESLA
intel inside
HD Graphics
Stratix 10
M2
CPUs
GPUs
Integrated GPUs
FPGAs

Enabling Acceleration for Managed Runtime Languages



Enabling Acceleration for Managed Runtime Languages



TornadoVM

Parallel Programming Framework

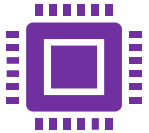
for

Accelerating

Java Data Parallel Workloads

on

Heterogeneous Hardware



Parallel Programming API

Three JIT Compilers

- *OpenCL C*
- *CUDA PTX*
- *SPIR-V*

Optimizer

- *100+ optimization phases*

Optimising Runtime System

Runs on JDK 21 & JDK 23

Dynamic Task Reconfiguration

Muti-device

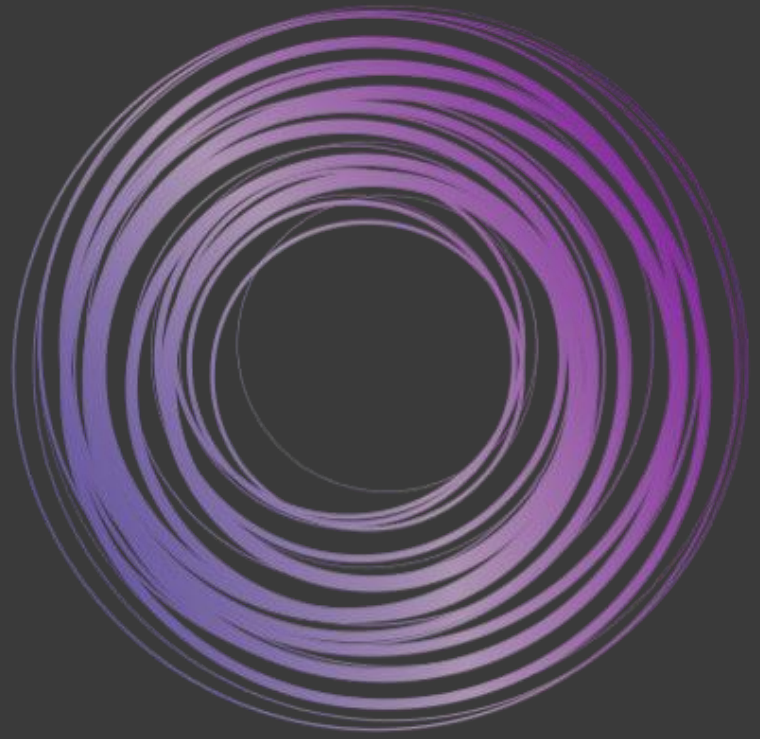
Automatic Batch Processing

Automatic Data Management

Multi-backend

and it is **Open Source**:

https://github.com/beehive-lab/TornadoVM



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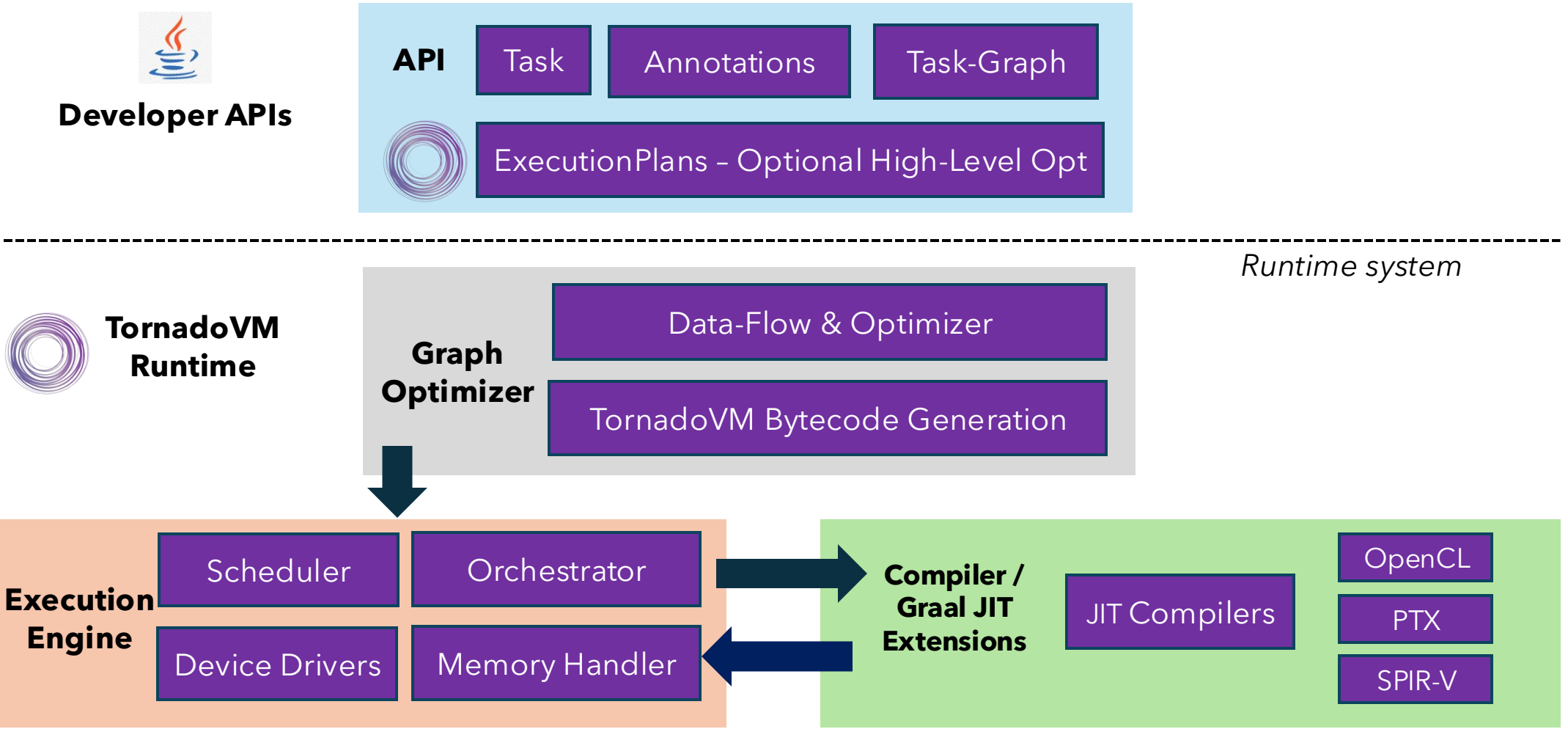
TornadoVM APIs

TornadoVM's Software Stack



Different components of the User APIs

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a) **How to represent parallelism within functions/methods?**
   - A.1: Java annotations for expressing parallelism (**@Parallel, @Reduce**) for Non-Experts
   - A.2: Kernel API for GPU experts (use of **kernel context** object)

b) **How to define which methods to accelerate?**
   Build a **Task-Graph API** to define data In/Out and the code to be accelerated

c) **How to explore different optimizations?**
   Execution Plan

Let's Learn the Different API Components with an Example

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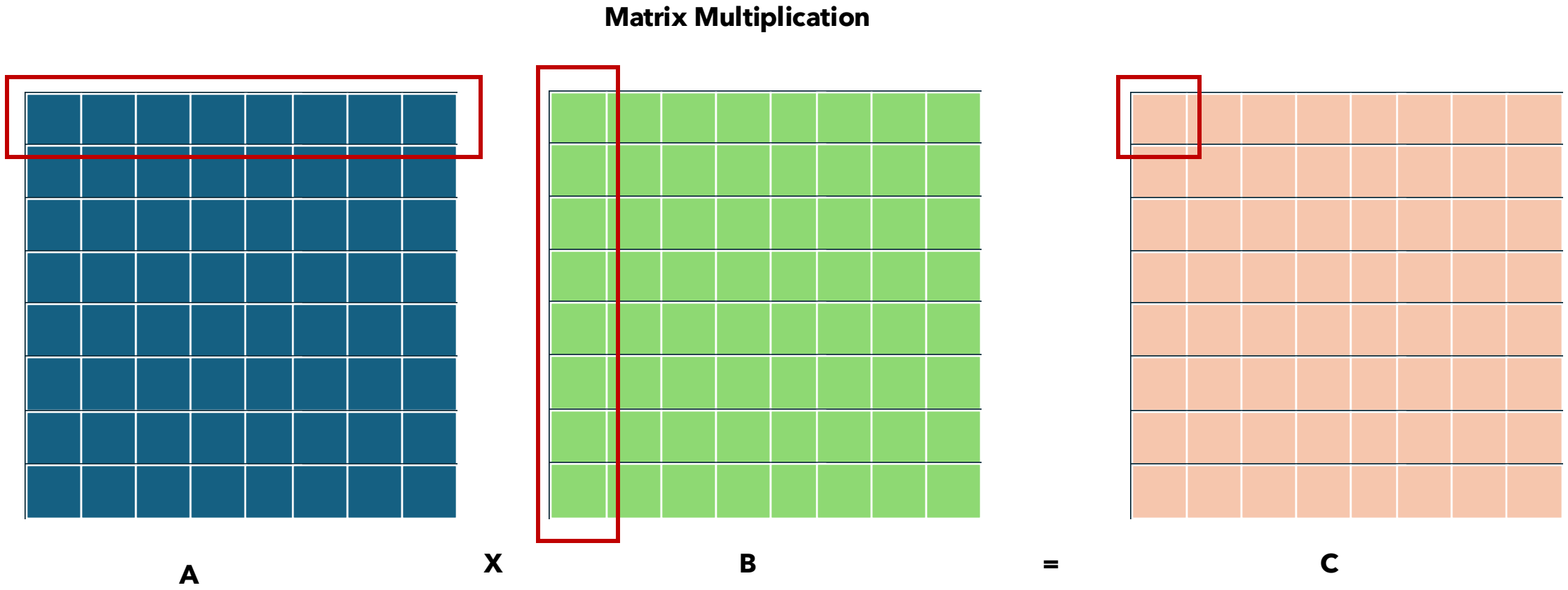

- Widely used on ML and AI
- Linear Algebra Kernel

https://github.com/beehive-lab/TornadoVM/blob/master/tornado-examples/src/main/java/uk/ac/manchester/tornado/examples/compute/MatrixMultiplication2D.java

Example using the TornadoVM Loop Parallel API

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```
class Compute {
  public static void mxm(Matrix2DFloat A, Matrix2DFloat B,
                         Matrix2DFloat C, final int size) {
     for (int i = 0; i < size; i++) {
        for (int j = 0; j < size; j++) {
           float sum = 0.0f;
           for (int k = 0; k < size; k++) {
              sum += A.get(i, k) * B.get(k, j);
           }
           C.set(i, j, sum);
        }
     }
  }
}
```

https://github.com/beehive-lab/TornadoVM/blob/master/tornado-examples/src/main/java/uk/ac/manchester/tornado/examples/compute/MatrixMultiplication2D.java

Tornado API – example using Annotations

Panama-Based Memory Segments for Off-Heap Data (*)

```
class Compute {
  public static void mxm(Matrix2DFloat A, Matrix2DFloat B,
                         Matrix2DFloat C, final int size) {
     for (@Parallel int i = 0; i < size; i++) {
        for (@Parallel int j = 0; j < size; j++) {
           float sum = 0.0f;
           for (int k = 0; k < size; k++) {
              sum += A.get(i, k) * B.get(k, j);
           }
           C.set(i, j, sum);
        }
     }
  }
}
```

We add the `@Parallel` Annotation

We add the parallel annotation as a hint for the compiler

We only have 2 annotations:

@Parallel

@Reduce

Tornado API – example using Kernel Context

```
class Compute {
  public static void mxm(Matrix2DFloat A, Matrix2DFloat B,
                         Matrix2DFloat C, final int size,
                         KernelContext context) {
      int idx = context.globalIdx;
      int jdx = context.globalIdy;
      float sum = 0.0f;
      for (int k = 0; k < size; k++)
         sum += A.get(idx, k) * B.get(k, jdx);
      C.set(idx, jdx, sum);
  }
}
```

Kernel-Context accesses thread ids, local memory and barriers

It needs a **Grid of Threads** to be passed during the kernel launch

Alternative API for expert programmers. It offers more control

Tornado API – example

How to identify which methods to accelerate? --> **TaskGraph**

```
TaskGraph taskGraph = new TaskGraph("myComputeGraph")
    .transferToDevice(DataTransferMode.EVERY_EXECUTION , matrixA, matrixB)
    .task("parallelTask", Compute::mxm, matrixA, matrixB, matrixC, size)
    .transferToHost(DataTransferMode.EVERY_EXECUTION, matrixC);
```

Host Code

TaskGraph is a new TornadoVM object exposed to developers to define :

a) **The code to be accelerated** (which Java methods?)

b) **The data (Input/Output)** and how data should be streamed

Defining and Running Execution Plans

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How to run/explore different optimizations? --> `ExecutionPlan`

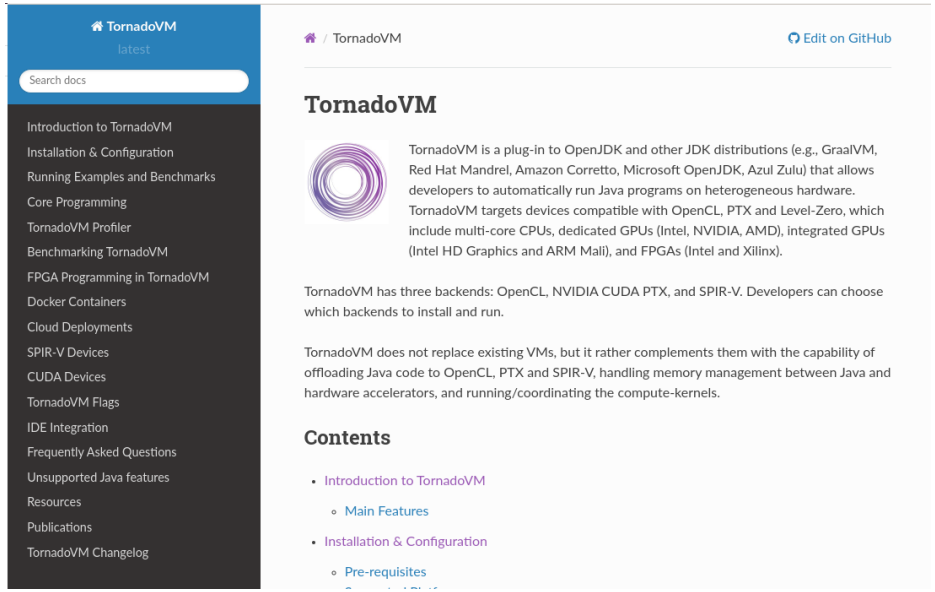
```
TornadoExecutionPlan executionPlan = new TornadoExecutionPlan(taskGraph.snapshot());

// optional: change runtime parameters/optimize the runtime
executionPlan.withWarmUp()
             .withProfiler(...)
             .withDynamicReconfiguration(PERFORMANCE, ...);

// blocking call: the execute is where JIT compilation + execution happen
TornadoExecutionResult result = executionPlan.execute();

long elapsedKernelTime = result.getProfiler().getDeviceKernelTime(); // in ns
```

Host Code

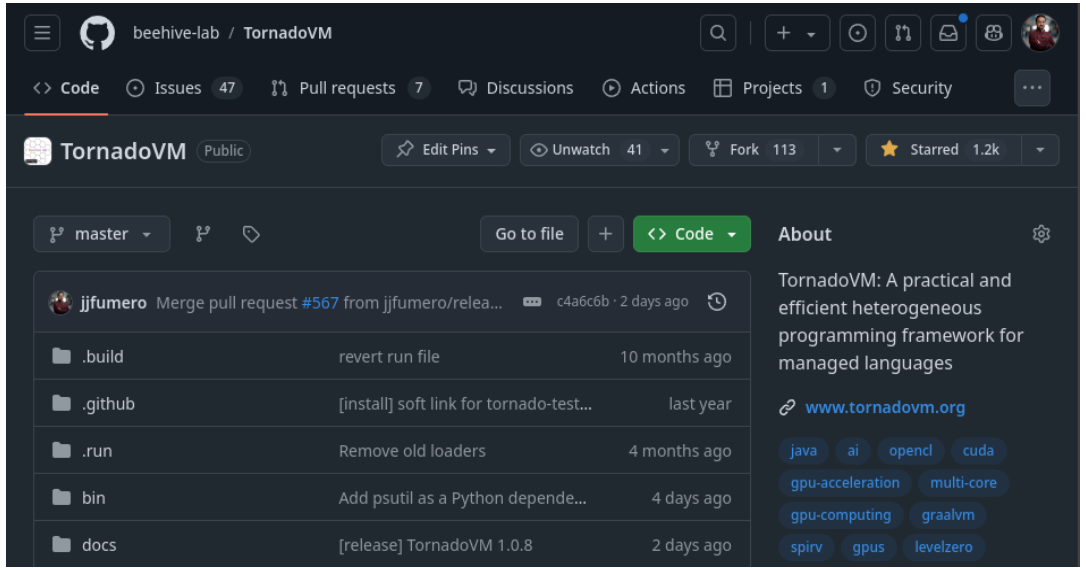
Optional High-Level Optimization Pipelines:

- Enable/Disable Profiler
- Enable Warmup
- Enable Dynamic Reconfiguration
- Enable Batch Processing
- Enable Thread Scheduler (no need for recompilation for different grids schedulers)

To learn more about the APIs



https://tornadovm.readthedocs.io/en/latest/



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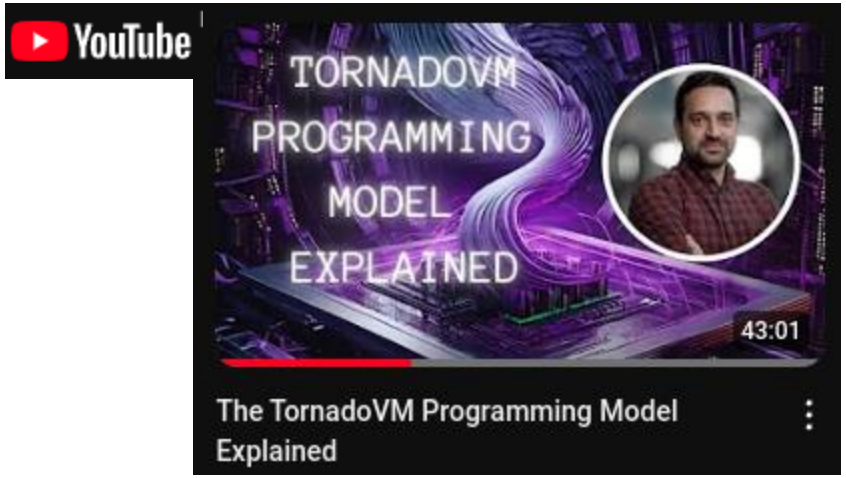

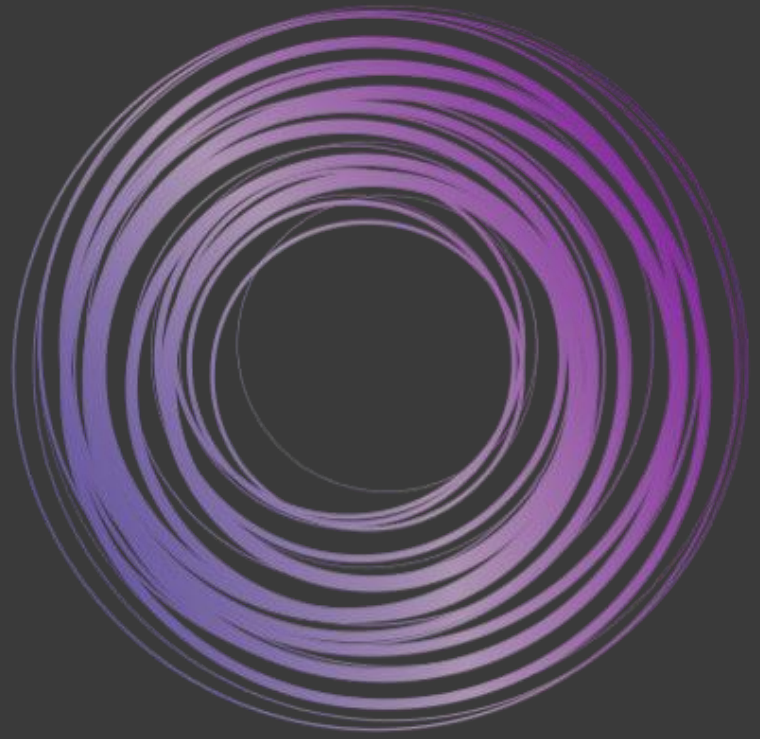

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How TornadoVM can be used on RISC-V for RVV 1.0?

oneAPI Construction Kit (OCK)

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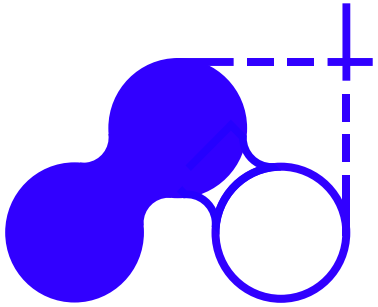
https://github.com/uxlfoundation/oneapi-construction-kit



"Programming Framework to enable hardware platforms to access open standards (e.g., SYCL, OpenCL, etc) and it can be used to extend the oneAPI software ecosystem to custom compute architectures [1]".

UXL
Unified Acceleration Foundation

Part of the UXL Foundation: https://uxlfoundation.org/

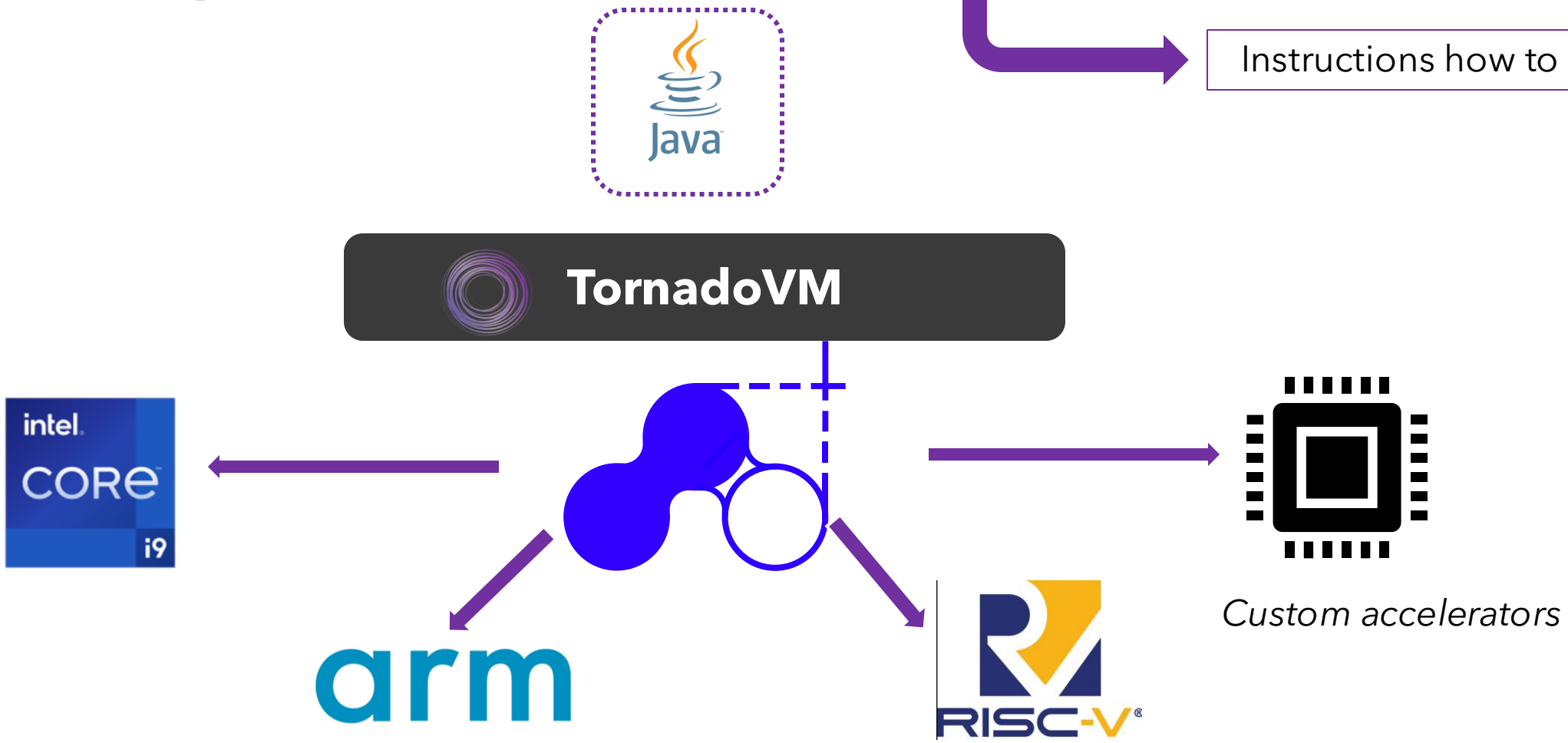
Working with Codeplay in the AERO EU Project [2]

[1] https://developer.codeplay.com/products/oneapi/construction-kit/home/
[2] https://aero-project.eu/

Running on Multi-Vendor CPUs: ARM, RISC-V, Intel
Using the **oneAPI Construction Kit (OCK)**

https://jjfumero.github.io/posts/2024/09/10/tornadovm-ock

Instructions how to build



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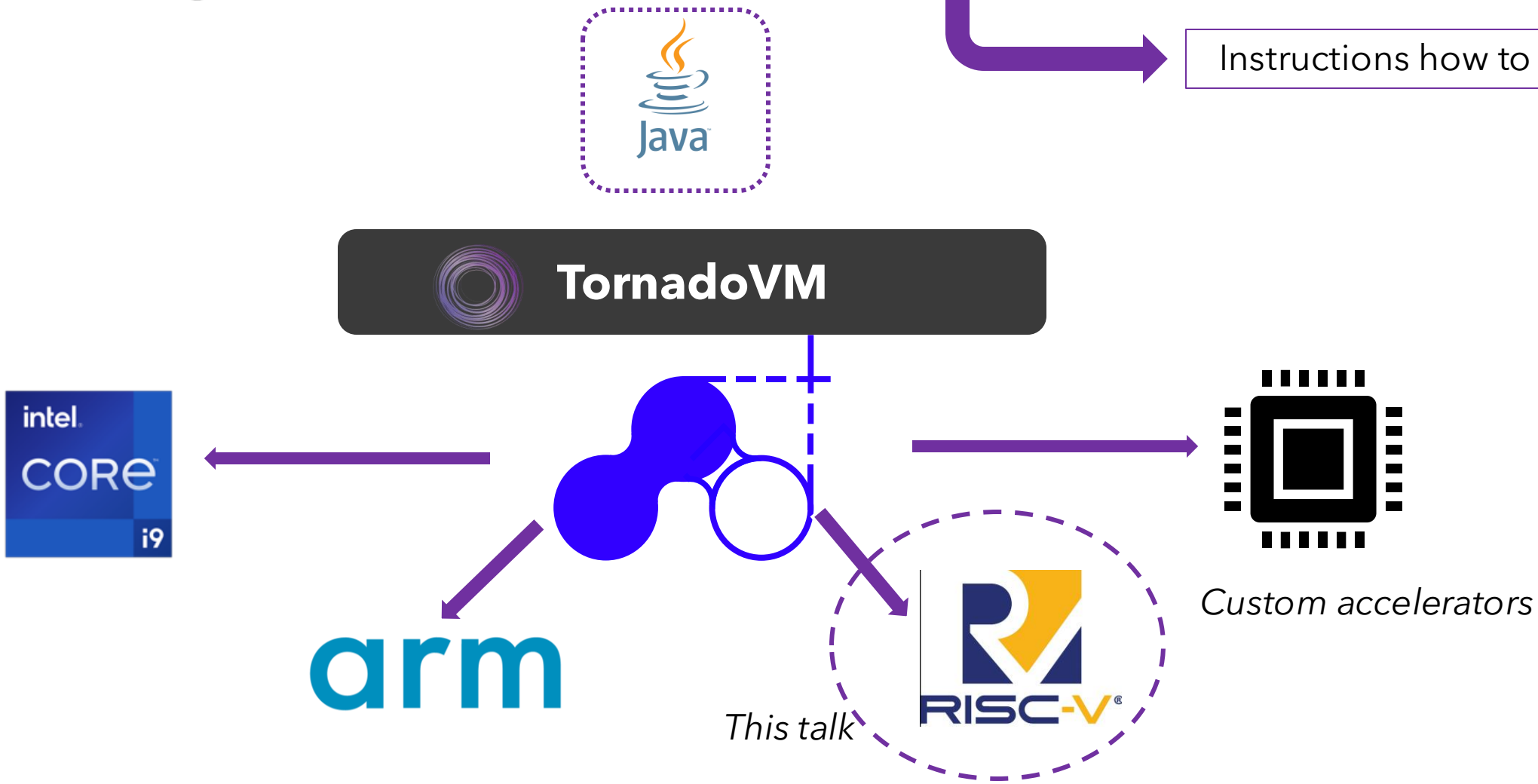
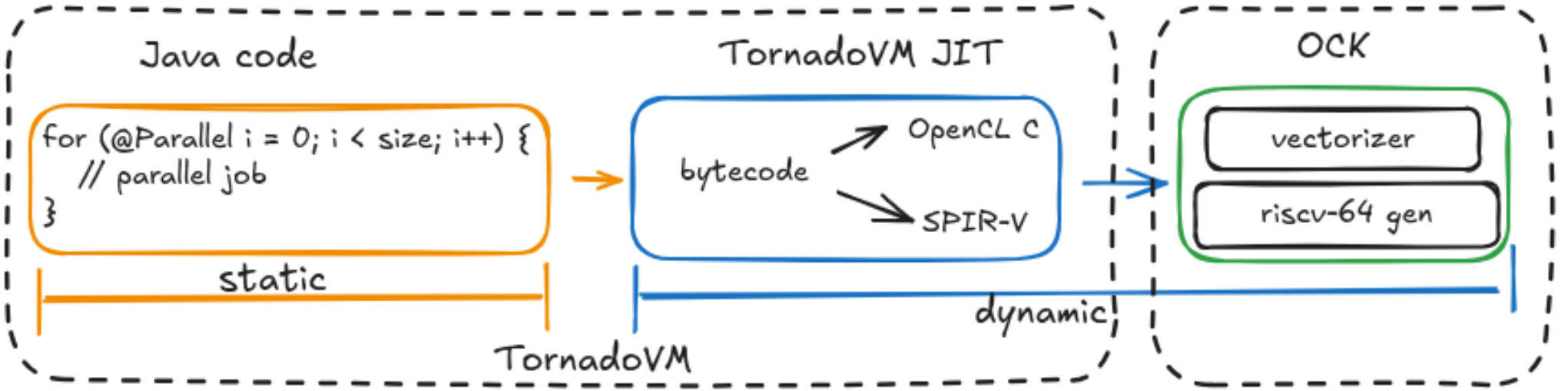
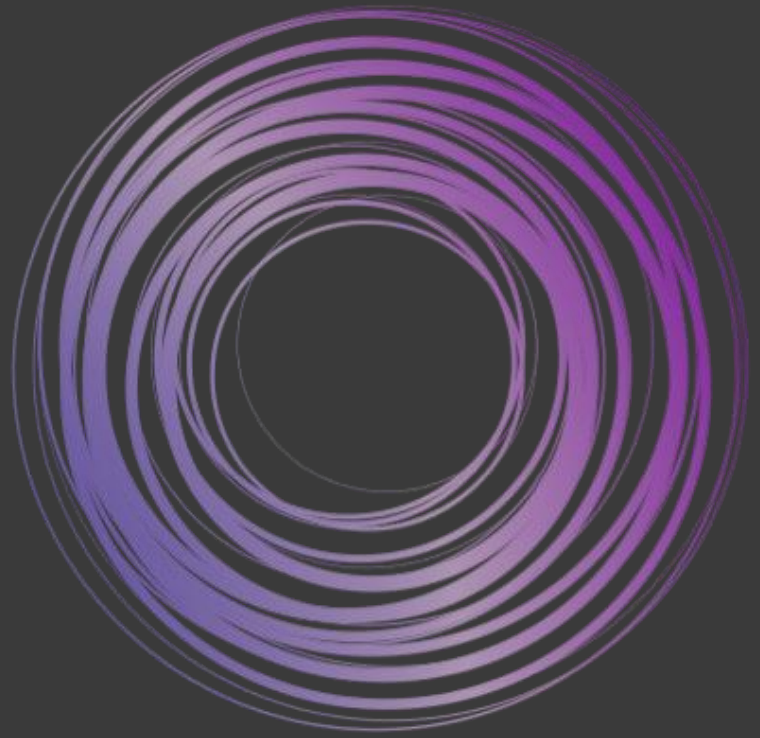


JIT Compilation Process



TornadoVM contains more than 50 compilation phases
Compilation is per-device, per architecture. This is crucial to get performance portability



TORNADO VM

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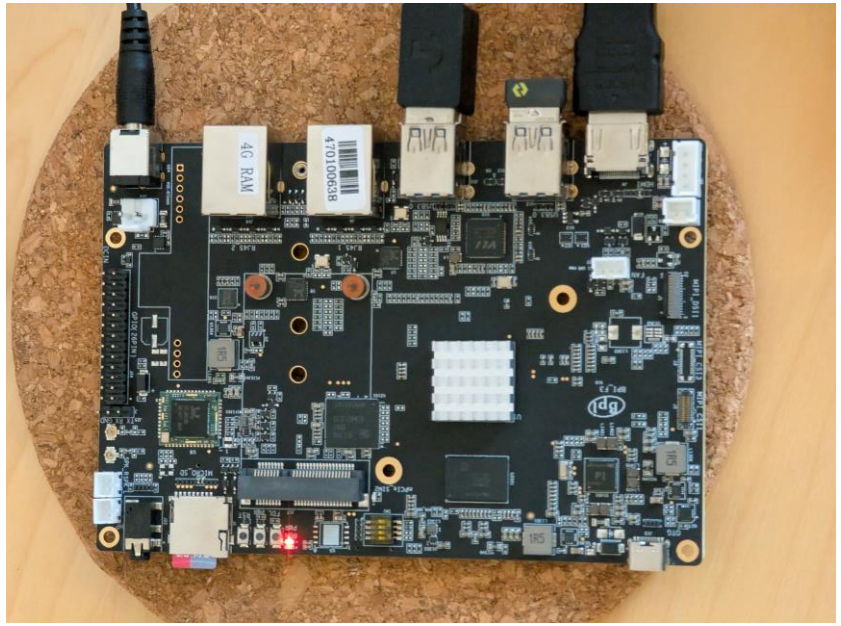
Performance

SBC used

RISC-V Spacemit K1 Processor

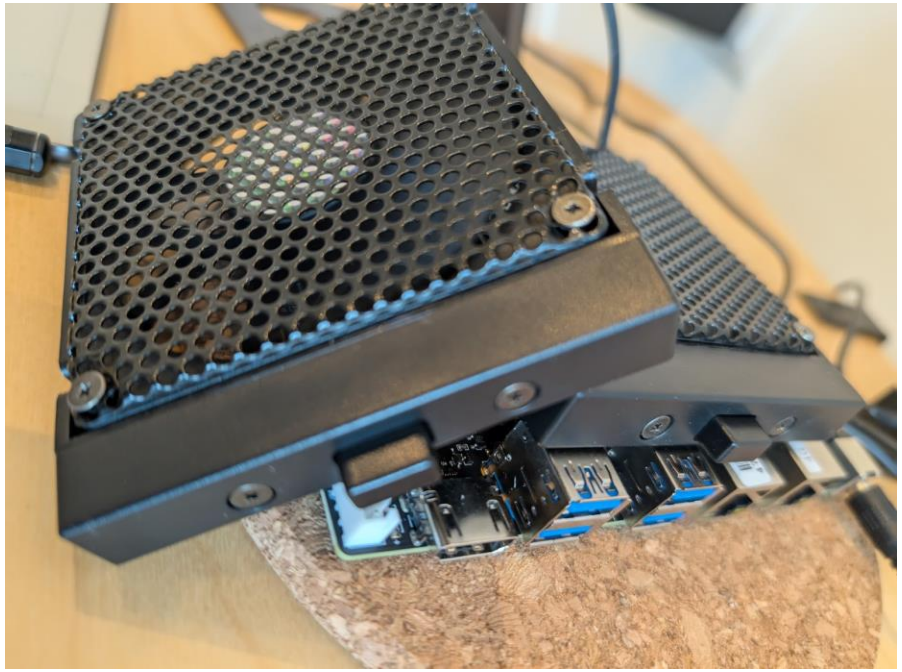
- Banana PI BPI F3
- Sipeed Lichee PI 3A

Octacore RISC-V @ 1.6GHz
RVA22 and 256-bit RVV 1.0
1MB shared L2 Cache
Max 16GB of RAM

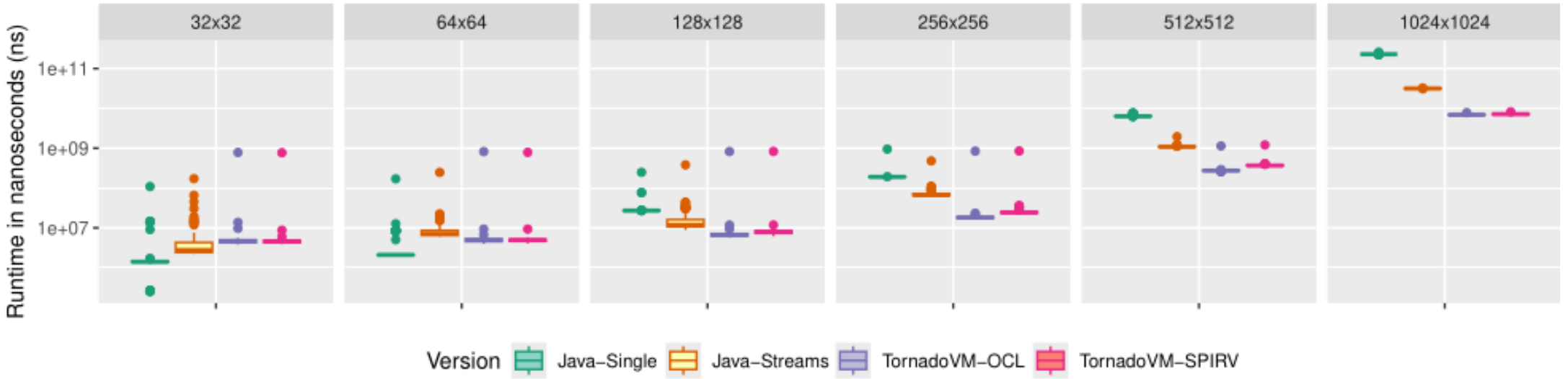


Things to be aware of

- Need to be extremely patient. This development board is extremely slow.
- ~4 days to compile LLVM on the RISC-V
  - Single thread compilation to avoid swapping
- ~2 days to compile OCK
  - Multiple configurations:
    - ***Refsi*** (RISC-V Emulator on RISC-V)
    - No vectorization
    - **With auto-vectorization**
- Use active cooling to avoid throttling



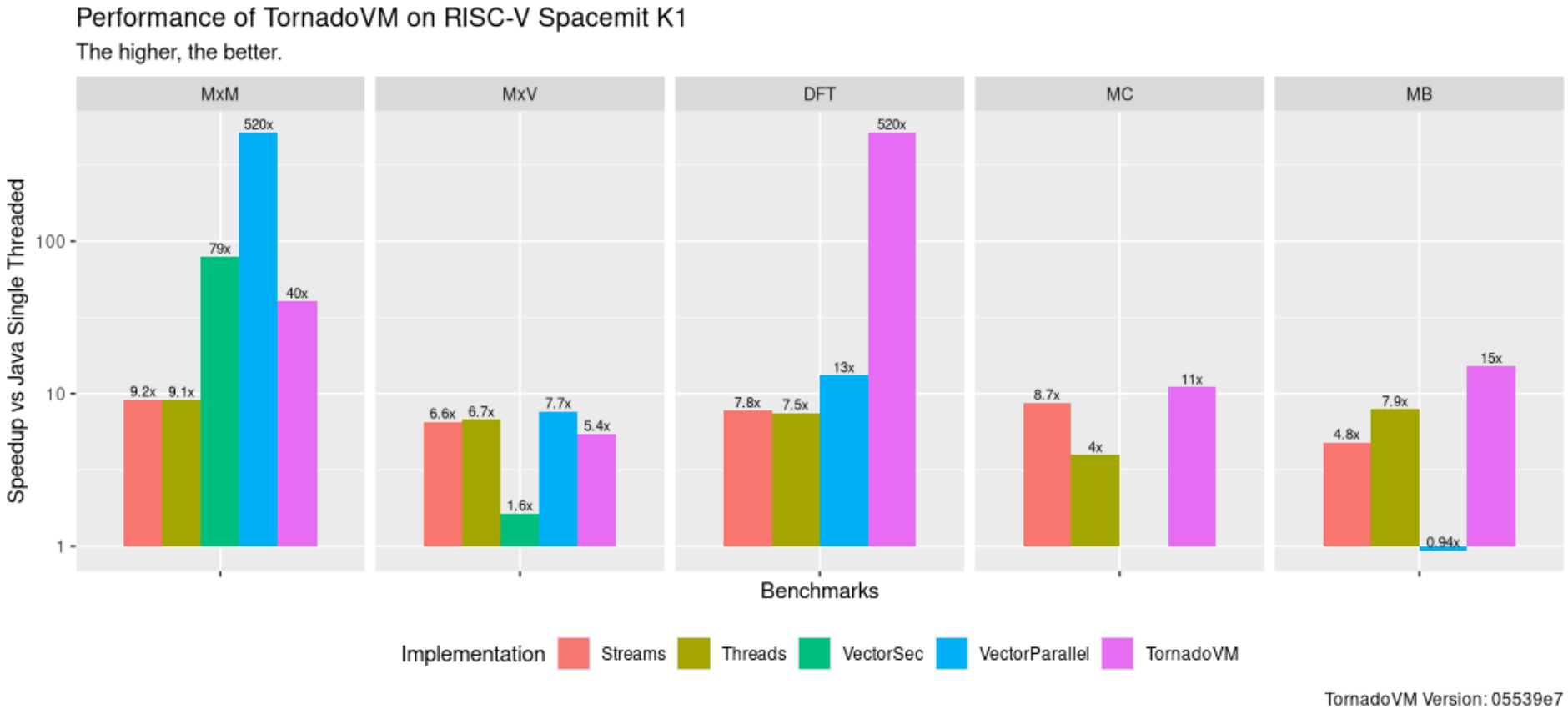
Performance of MxM on RISC-V (Banana PI 4GB)



TornadoVM 1.0.10-dev (ec667bd65)
LLVM 19.1.5
GCC 13.2
OpenJDK 21.0.5
Banana PI F3

- For small matrices, stay on single core.
  - No auto-vectorization for Java on RISC-V as in FEB 2025 (we checked the generated Assembly code)
- Java streams is 2.3x-7.2x faster than single core for larger matrices
- TornadoVM + OCK achieves 4.1x-33x faster than Java seq, and up to 4.6x vs Streams
- SPIR-V backend is bit slower compared to OpenCL C

Performance on RISC-V Spacemit K1 (Lichee PI3A)



Explicit Vector API is quite competitive
Hard to express Vector API in some applications

TornadoVM: 0553ae7 (1.0.11-dev)
LLVM 19.1.5
GCC 13.2
OpenJDK 21.0.5
Lichee PI 3A

Vector API uses Intrinsics

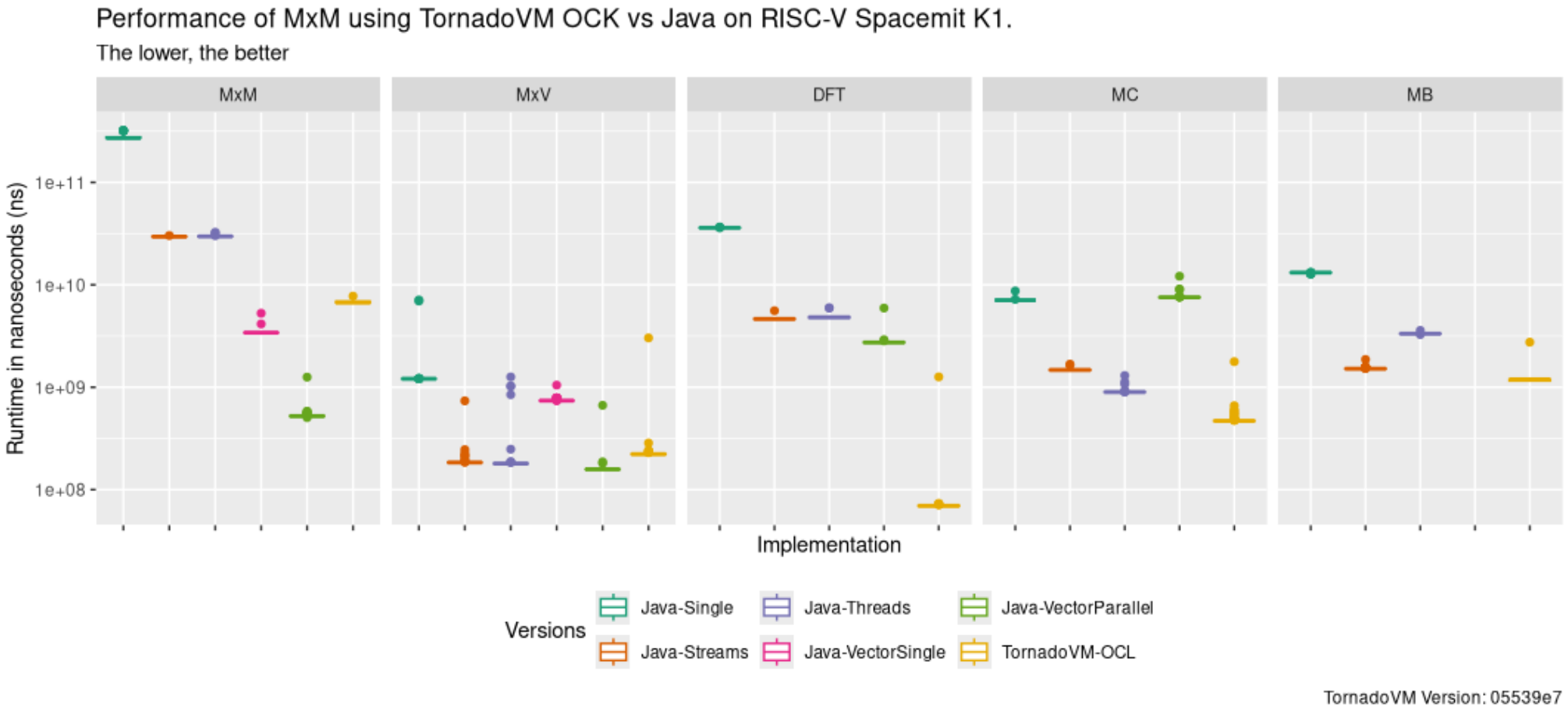
Java JIT Compiled Code (C2), using `-XX:+UnlockDiagnosticVMOptions -XX:+PrintAssembly`

```
0x00007f469cdb7a0f:   cmp    %r11,%r10
0x00007f469cdb7a12:   jne    0x00007f469cdb8ad4
0x00007f469cdb7a18:   mov    %rcx,%rax   ;*invokestatic reductionCoerced {reexecute=0 rethrow=0 return_oop=0}
                                         ; - jdk.incubator.vector.FloatVector::reduceLanesTemplate@78 (line 2644)
                                         ; - jdk.incubator.vector.Float256Vector::reduceLanes@2 (line 324)
                                         ; - tornadovm.benchmarks.DFT::lambda$computeWithParallelVectorAPI$1@263 (line
  136)
```

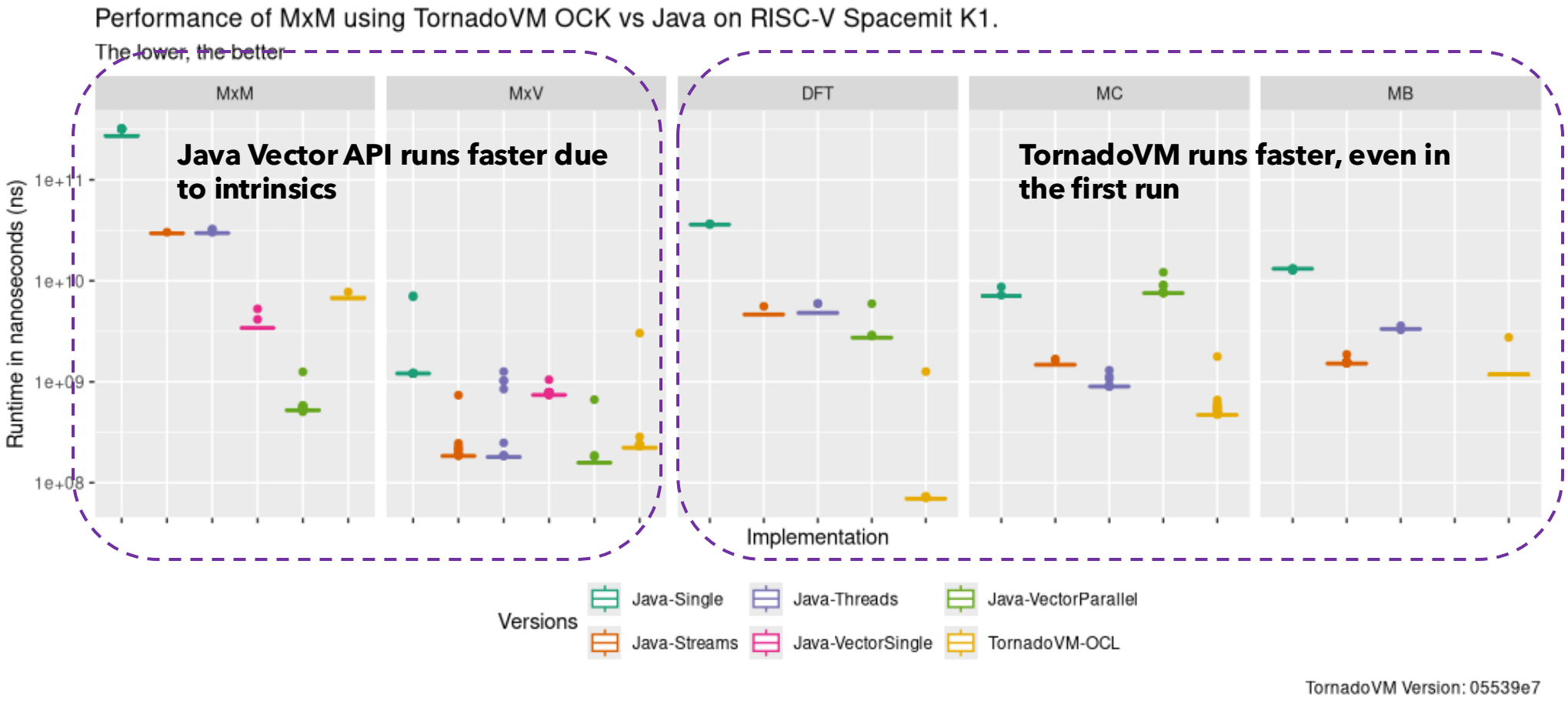
Java Source Code:

```
136  sumimag += -1 * vInReal.mul(FloatVector.fromArray(species, sinAngles, 0))
                   .add(vInImag.mul(FloatVector.fromArray(species, cosAngles, 0)))
                   .reduceLanes(VectorOperators.ADD);
```

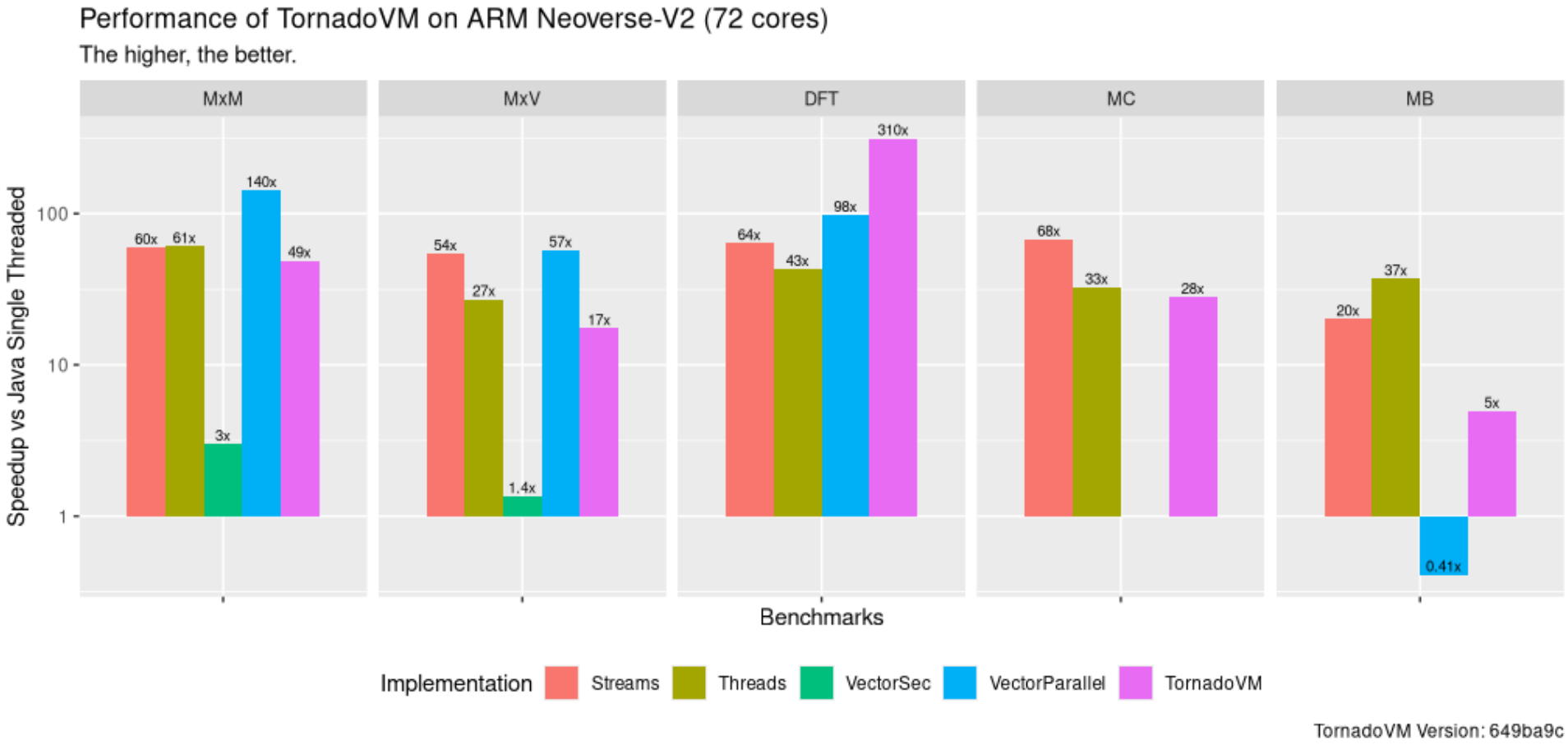
Runtime Distribution on RISC-V Spacemit K1 (Lichee PI3A)



Runtime Distribution on RISC-V Spacemit K1 (Lichee PI3A)



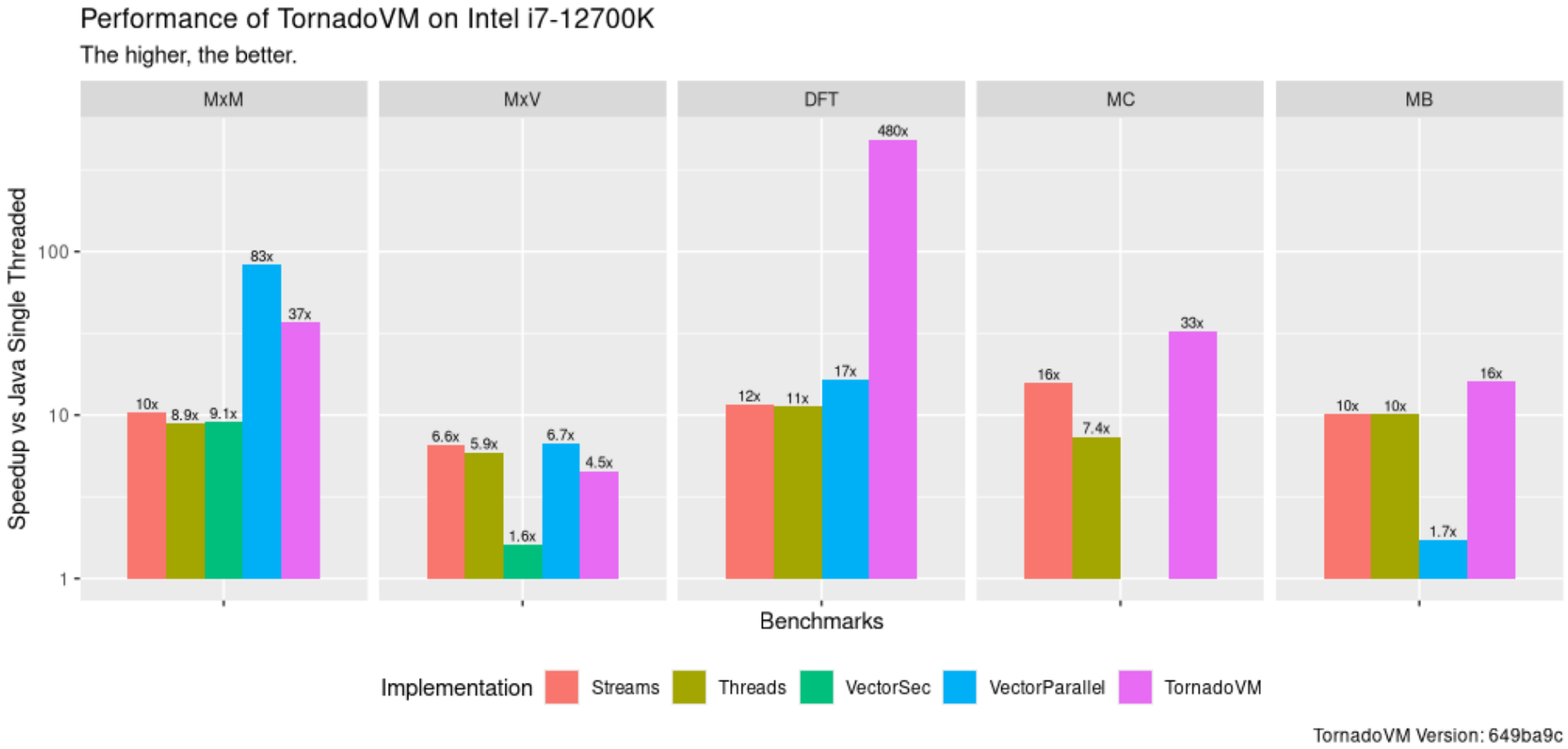
[**ARM CPUs**] ARM Neoverse V2 (Grace Hopper)



ARM Neoverse V2 CPU (NVIDIA Grace Hopper)
Ubuntu 22.04.4 LTS
Kernel 6.2.0-1015-nvidia-64k
OCK 4.0: 5be5a8d
TornadoVM 1.0.11-dev: 649ba9c
OpenJDK 21.0.3+7-LTS

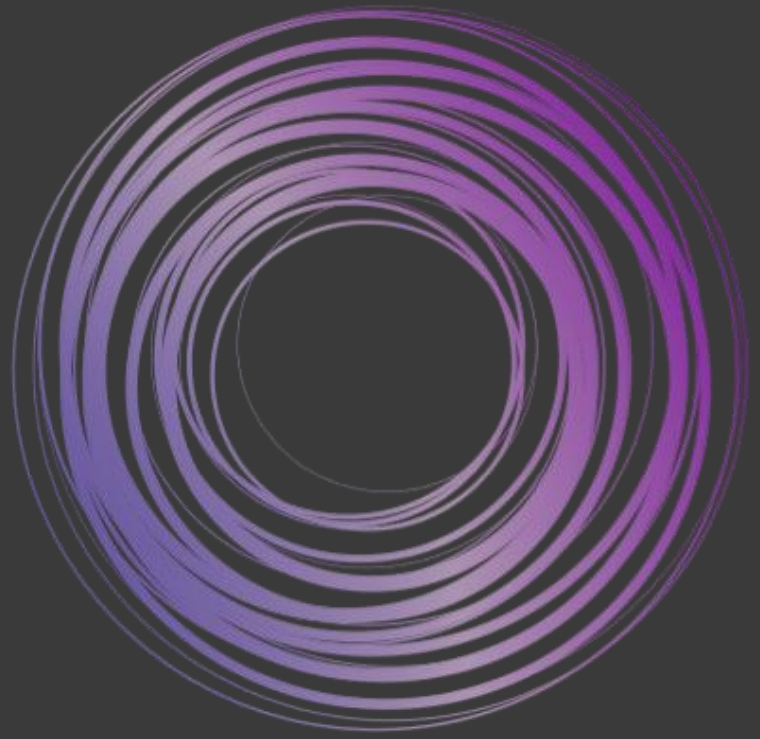
Java multi-threaded is very competitive in this platform, since OCK does not use SVE instructions. Explicit vectorization is a mixed: if auto-vectorization works, there is no major benefit, at least for these benchmarks.

[**INTEL CPUs**] TornadoVM + OCK/oneAPI



Intel i7-12700K
Fedora 41
Kernel 6.2.15-200
OCK 4.0: a537ec99
TornadoVM 1.0.11-dev: 649ba9c
OpenJDK 21.0.6

For Intel CPUs, the Java JIT compiler generates vector instructions
TornadoVM + OCK generates efficient code just for the kernel of interest



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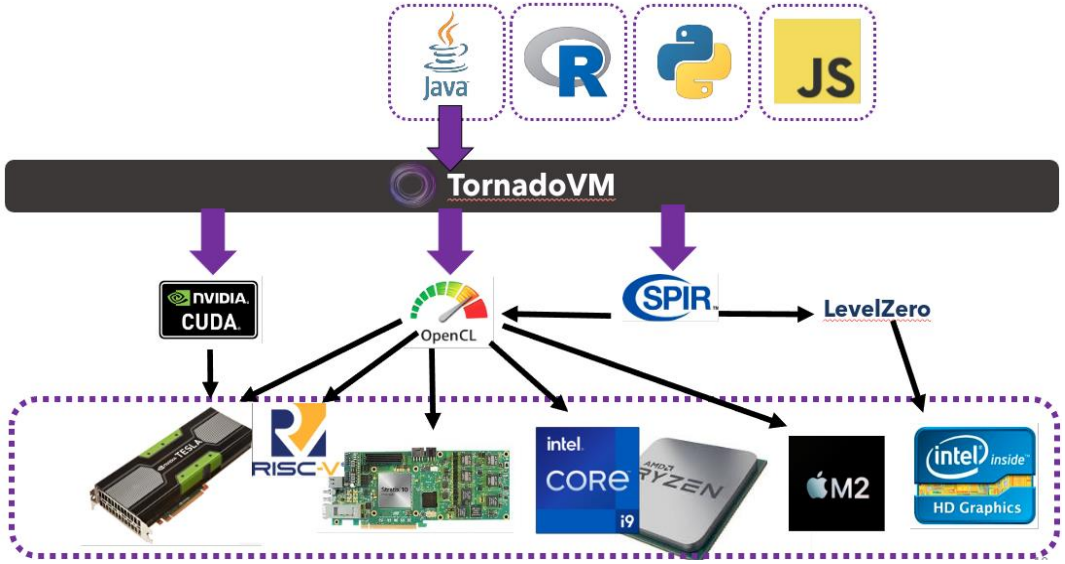
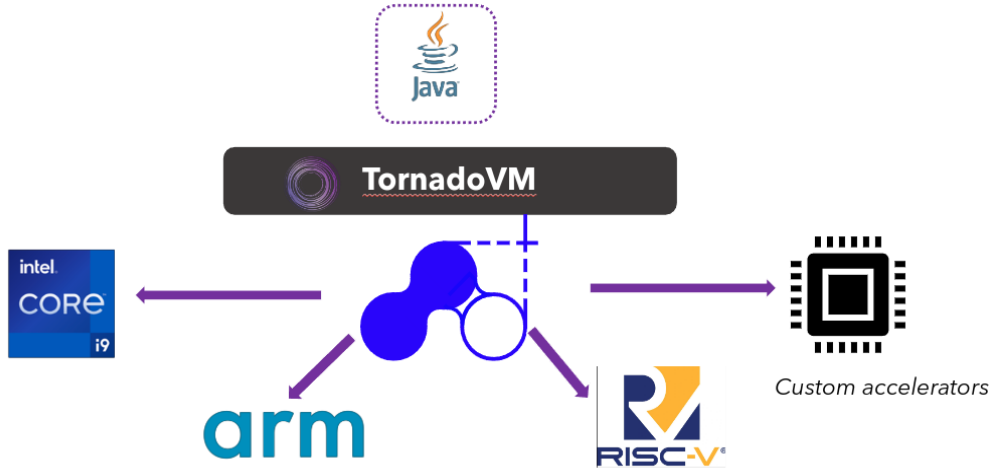
Conclusions

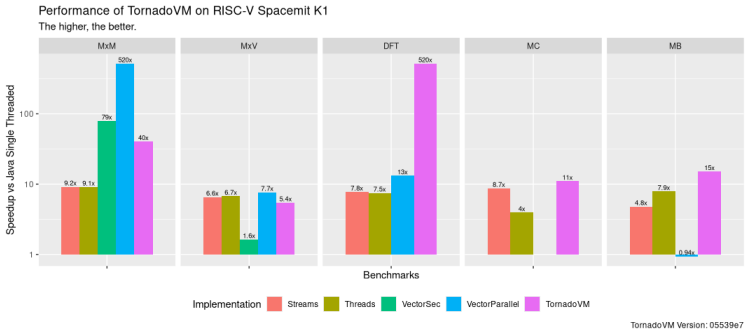
Key Takeaways





Faster than Java multi-core

Usually slower than explicit Java vectorization



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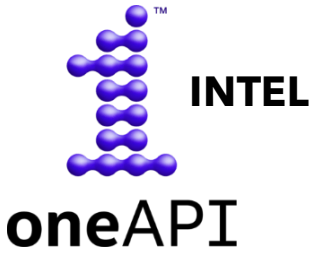
Collaborations and Projects

Project Number: 101092850

Project Number: 101093069

Project Number: 101070670

Project Number: 101070052



Thank you!

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Discussions