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Vector API design: Selected topics

(flat values, future shapes, cross-lane motion, snowflakes, atomics)

John Rose, Oracle JPG http://cr.openjdk.java.net/~jrose/pres/202202-VectorTopics.pdf Feb 16, 2022

Fitting vectors into bare values — a few ideas

Key observation: Vector lanes are indexed, not selected by name

- A vector is a homogeneous short array
 - A value class is a heterogenous tuple
- Array indexing v[i] requires an in-memory presentation
 - A value class, even if in memory, cannot index its fields
 - · Class fields are often, but not always, contiguous
- First compromise: A private multi-field feature in the class loader
 - One field, secretly replicated N times CONTIGUOUSLY
 - (Probably) allocated by JVM after any/all other fields
 - Access (other than to first copy) is via var-handle only
- Additional compromise: Register allocator assigns it a vector register
 - Either automatic, or an extra ad hoc register class annotation

More details: <u>http://cr.openjdk.java.net/~jrose/values/multi-field.html</u>

Example "multi-field" feature

```
@RegisterAllocation(128) value class Shape128 {
   @MultiField(2) long data;
   //long data#1 ← allocated contiguously
@RegisterAllocation(256) value class Shape256 {
   @MultiField(4) long data;
   //long data#{1,2,3} ← allocated contiguously
@RegisterAllocation(512) value class Shape512 {
   @MultiField(8) long data;
   //long data#{1,2,3,4,5,6,7} ← allocated contiguously
```



Example var-handle based access (HALF-BAKED)

```
value class Shape512 {
   @MultiField(8) long data;
   private static long OFF_data = U.objectFieldOffset(F_data);
   private static final VarHandle VH_data =
     new VarHandleMultiLongs(Shape512.class, OFF_data, 8);
   public long data(int i) {
     return (long) VH_data.get(this, i);
   }
   public Shape512 dataUpdate(int i, long x) {
     Shape512 pbuf = U.makePrivateBuffer(this);
     VH_data.set(pbuf, i, x);
     return U.finishPrivateBuffer(pbuf);
   } //OR return (Shape512) VH_data.'update'(this, i, x);
```

After discussion, the 'update' operator doesn't look so good; use arrays

```
value class Shape512 {
   @MultiField(8) long data;
   private static long OFF_data = U.objectFieldOffset(F_data);
   private static final VarHandle VH_data =
     new VarHandleMultiLongs(Shape512.class, OFF_data, 8);
   public long data(int i) {
     return (long) VH_data.get(this, i);
   }
   public static Shape512 make(long[] a) { //optimizer makes temporary 'a' disappear
     Shape512 pbuf = U.makePrivateBuffer(this);
     for (int i = 0; i < 8; i++)
       VH_data.set(pbuf, i, a[i]);
     return U.finishPrivateBuffer(pbuf);
```

Straightforward to expand to multi-vector shapes

```
@RegisterAllocation(128) value class Shape128x2 {
   @MultiField(2*2) long data;
@RegisterAllocation(256) value class Shape256x3 {
   @MultiField(4*3) long data;
@RegisterAllocation(512) value class Shape512x4 {
   @MultiField(8*4) long data;
// Nested representation would be more complex to implement
value class Shape256x5 {
   @MultiField(5) Shape256 nestedVector;
   // requires carefully contiguous layout of the nestedVectors
```

Idea from the group: Maybe have "one size fits most" for polymorphic?

@RegisterAllocation(128) value class Shape128 extends Shape { @MultiField(2) long data; }
@RegisterAllocation(256) value class Shape256 extends Shape { @MultiField(4) long data; }
@RegisterAllocation(512) value class Shape512 extends Shape { @MultiField(8) long data; }

// AND THEN:

@RegisterAllocation(512 /*=MAX*/)
abstract class Shape permits Shape128, Shape256, Shape512 { }

// ... models the overlapping zmm/ymm/xmm register file
// with a klass token, maybe useful for polymorphic calling conventions??



The shapes of things to come —rectangles, tiles, matrices

Key observation: Multi-vector algorithms are "a thing"

- Logically rectangular "tiles" (as in the new Intel AMX feature)
 - Show up in optimized "BLAS microkernels" (as in BLIS, GotoBLAS).
 - C[m,n] += Sum[k] A[m,k] * B[k,n] for m,n small, like lane counts (4/6/8)
 - Key operations: Outer product (A[*,k] against B[k,*]), then add into C[*,*] for each k
 - Motion along either logical dimension: replicate, reduce, reorder, transpose
- · Loop unrolling: Use "as many registers as you can get away with"
 - Requires testing to see how many that might be (playing "The Price is Right" in vector register file)
 - Easier to configure as a multi-vector shape, not "unroll and jam" of source code
 - This technique combines with tiles: Keep adding tiles until we run out of registers
 - Common operation: Partial reduction (N-way unroll \Rightarrow N-to-1 reduction of accumulator)
- Vector-of-structs: A vector of three-element structs should be three vectors.
 - Common operation: Transpose (SoV ⇔ VoS), other ad hoc shuffles.

Some basic operations for multi-vectors

- Native vector species (VLEN1) replicated by some small RLEN, VLEN2=VLEN1 x RLEN
- All the usual lanewise stuff. Also slicing, compression, (full) reduction. Masking (with big masks).
- Shuffle (with a jumbo index array)
 - Some set patterns like transpose, zip/unzip, SoV / VoS
 - Support partial broadcast (a,b,c...) to (a,a,a,b,b,b,c,c,c...) or (a,b,c...,a,b,c...,a,b,c...)
 - Compiles to RLEN single-vector shuffles, possibly with RLEN merges.
 - Partial broadcast can set up an outer product (matrix tile multiply/FMA)
- Reshaping shuffle (expansion/contraction) is probably important
 - · Generalizes today's pick-from-two-vectors shuffle operation
 - Use to extract or insert or broadcast VLEN1-vector against VLEN2 vector
- Striding partial reduction = (a,b,c,d,e,f) to (a+c+e, b+d+f) (reduce by RLEN)
 - Other orders of reduction are best obtained by a previous shuffle?



Challenge problem: BLIS microkernel







Challenge problem: BLIS microkernel



The Bliss SGEMM microkernel for Haswell is here: <u>https://github.com/flame/blis/blob/HEAD/kernels/haswell/3/bli_gemm_haswell_asm_d6x8.c#L177</u>

// C<96> += A<16> * B<6> // [ymm4..ymm15]<96> += [ymm0..ymm1]A<16> * [(ymm2..ymm3)x3]B<6> vbroadcastss(mem(rbx, 6*4), ymm2) vbroadcastss(mem(rbx, 7*4), ymm3) vfmadd231ps(ymm0, ymm2, ymm4) vfmadd231ps(ymm1, ymm2, ymm5) vfmadd231ps(ymm0, ymm3, ymm6) vfmadd231ps(ymm1, ymm3, ymm7) vbroadcastss(mem(rbx, 8*4), ymm2) vbroadcastss(mem(rbx, 9*4), ymm3) vfmadd231ps(ymm0, ymm2, ymm8) vfmadd231ps(ymm1, ymm2, ymm9) vfmadd231ps(ymm0, ymm3, ymm10) vfmadd231ps(ymm1, ymm3, ymm11) vbroadcastss(mem(rbx, 10*4), ymm2) vbroadcastss(mem(rbx, 11*4), ymm3) vfmadd231ps(ymm0, ymm2, ymm12) vfmadd231ps(ymm1, ymm2, ymm13) vfmadd231ps(ymm0, ymm3, ymm14) vfmadd231ps(ymm1, ymm3, ymm15)

Challenge problem: BLIS microkernel



```
// C<96> += A<16> * B<6>
// [ymm4..ymm15]<96> += [ymm0..ymm1]A<16> * [(ymm2..ymm3)x3]B<6>
```

```
var A_SP = FloatVector.SPECIES_256.times(2); //<16>
var B_SP = FloatVector.SPECIES_64.times(3); //<6>
var C_SP = FloatVector.SPECIES_256.times(12); //<96>
var C_V = FloatVector.fromArray(C_SP, C_ARR, cIndex); cIndex += 96;
var A_V = FloatVector.fromArray(A_SP, A_ARR, aIndex); aIndex += 16;
var AREPL_V = C_V.broadcastCyclic(A_V);
for (;;) { ...
var B_V = FloatVector.fromArray(B_SP, B_ARR, bIndex); bIndex += 6;
var BSPREAD_V = C_V.broadcastExpand(B_V);
C_V = C_V.fma(AREPL_V, BSPREAD_V);
```

Challenge accepted: Paul Sandoz writes it w/o multi-vectors

https://mail.openjdk.java.net/pipermail/panama-dev/2021-March/012664.html

```
// Row 0: AB += OldC x beta
AxB_0L = vOldCL.fma(vbr_beta, AxB_0L);
AxB_0R = vOldCR.fma(vbr_beta, AxB_0R);
// Save results to free up registers:
AxB 0L.intoArray(c, c row iter);
```

```
AxB_0R.intoArray(c, c_row_iter + 8);
c row iter += rs c;
```

```
// Row 1: AB += OldC x beta
vOldCL = FloatVector.fromArray(SPECIES_256, c, c_row_iter);
vOldCR = FloatVector.fromArray(SPECIES_256, c, c_row_iter + 8);
AxB_1L = vOldCL.fma(vbr_beta, AxB_1L);
AxB_1R = vOldCR.fma(vbr_beta, AxB_1R);
AxB_1L.intoArray(c, c_row_iter);
AxB_1R.intoArray(c, c_row_iter + 8);
c row iter += rs c;
```

Cross-lane data motion —general observations

Kinds of cross-lane motion



- Random access shuffle: *Pulling* values to shuffle indexes which *attract* them.
 - · Completely general; use when all else fails or to simulate
- Sidelong slide: Slipping back and forth, *maintaining stable order* of active set
 - Compress, expand, *sheep-and-goats (bidirectional compress)
 - Useful for quick-sort pivot, collecting *conditional* loop outputs
 - Compress/expand is useful for parsing (expand regularizes token sizes)
- Sorting/binning: *Pushing* values to be consistent with *key order*
 - *Sort, can be simulated by 2lg(N) compress operations (doing radix sort)
 - Complementary to random access shuffle ("reverse data flow")
 - *Binning is sort followed by reduction of items with the same key
 - For such operations, a key vector drives reordering of a data vector
- Research problem: Co-design of software and hardware

Challenge problems: Parsing (using sidelong compress/expand)

- Load a block of UTF8 bytes
- Find the token boundaries (make a mask)
- Expand tokens to uniform size (24 or 32 bits)
- Perform bit-twiddling to normalize token to UTF16 values
- Compress to 16-bit lanes.
- Profit.

OR:

- · Load a block of ASCII characters.
- · Find natural language word boundaries.
- Expand tokens to uniform size (8 bytes, = max word size?)
- Strip out punctuation, white space; normalize case; visit the dictionary or histogram...

Appendix: What if my hardware doesn't do expand natively?

- Expand can be implemented as a by-index permutation (table lookup) plus a horizontal prefix-sum.
- First compute an index vector for the desired expansion.
- The index vector increases monotonically to the left, with a "bump" just after each mask position.
 - Example: Mask (0,0,1,0,0,1,1,0) derives the index vector (0,0,0,1,1,1,2,3).
 - Each index position sums up the total number of mask bits set to the left.
 - May require a series of lg(N) horizontal additions at various (power of two) offsets.
- Then perform a full index-based permutation, using the computed index vector to steer the input values.
 - The monotonically increasing index vector "pulls" each input value "to the right" into the output.



Appendix: OK, so what if my hardware doesn't do compress natively?

- Compress can be implemented as a masked indexed store (scatter) to a memory temporary.
- First compute an index vector for the desired compression.
- The index vector increases monotonically to the left, with a "bump" just after each mask position.
 - (Same as the index vector for compress, on the previous page!)
 - For unmasked positions, maybe store distinct, non-overlapping placeholder indexes.
 - Example: Mask (0,0,1,0,0,1,1,0) might derive the index vector (-8,-7,0,-6,-5,1,2,-4).
- Then perform a full index-based scatter, using the computed index vector to steer the input values.
 - The output buffer should be extra-sized, with half of the positions usable as "bit buckets" (-8 etc.).
 - If there's masked store primitive, just mask off the unused indexes, instead of using placeholders.
- After the memory finishes swirling around, pick up the assembled compressed output from the buffer.
- (Bonus trick: The placeholder indexes suggest how do masked scatters on AVX2 which lacks them.)

Challenge problems: Sorting (the Missing Primitive)

- Load a block of unsorted data
- Pick a pivot (maybe from this block, or maybe from a previous pass)
- Run bidirectional compress to separate across pivot (= Sheep-and-Goats from Hacker's Delight)
- Store the partitioned sub-blocks, or immediately re-sort (since it's in registers)
- When tired of sorting, store data for later merging

MUCH LATER...

- Load 2 or more merge blocks
- Perform an all-to-all (outer product) comparison
- Use the resulting bits to steer the merge

BOTH PHASES USE A SORTING OPERATION: How should this be coded?

Challenge problems: Taming the Permutation Zoo

- · Most permutations are "not very random"
- Is there some way to describe them concisely so that the compiler can select better instructions?
- Common permutations: zip/unzip, transpose, locally block-wise reordering, talk-to-neighbor (shift/slice)
- How to capture the regular ones? (The optimizer should "see through" shuffle expressions...)
- Connection to multi-vectors: Localized permutations can be expressed as less than R lg(R) shuffles
 - The general case requires lots of merge-ups (lg(R) merge tree), but the special ones are simple
 - The very common "partial broadcasts" (cyclic copy, expanding copy) have good locality
- Is there a better notation (transparent to optimizer), than an index vector, for the "easy" permutations?
- Data-driven shuffles (selectFrom lookups) should favor constant lookup tables.
 - Range or bitwise analysis on indexes should routinely remove range checks.
 - Recent case study: <u>https://mail.openjdk.java.net/pipermail/panama-dev/2022-February/016337.html</u>

RESEARCH PROBLEMS...

Catching snowflakes — getting to odd VPU instructions

Special data types

- Half-float (V)
- Bits128 (CLMUL, AES)
- Complex (many various component types!)
- These can probably be expressed as placeholder types: value objects
- The value object can have a single replicated field

```
value class ComplexFloat { @MultiField(2) float reim; }
value class ComplexDouble { @MultiField(2) double reim; }
value class Bits128 { @MultiField(2) long bits; }
value class Float16 { short bits; }
```



Special operations

- Carry-less multiply [p]clmul (V+S)
- aesenc/aesdec steps (V+S)
- vp2intersect: all-to-all (outer product) comparison with dual-dimension mask reduction
- vnni: 8/16b precision dot product, 32b precision accumulation (saturating!)
- AMX (matrix tiles); tile multiplication, dot product (like vnni)
- FMA, sometimes with odd parameter order or alternating add+sub variation
- Bit-permutation
 - PDEP, PEXT (S only, not V)
 - Reverse bits, reverse bytes, (rotate? transpose?)
- many ad hoc permutations and special loads (too many to keep track of)
 - vpermilps (simple butterfly-like shift within 128-bit mega-lanes); vperm2f128

Suggestion: Incubate snowflake intrinsics in JVM-only package

package jdk.internal.vm.data;

```
@RegisterAllocation(128) public value class Bits128 {
    private @MultiField(2) lo_hi;
    public static Bits128 make(long hi, long lo) { ... }
```

```
public static Bits128 bitwiseAnd(Bits128 a, Bits128 b) { ... }
public static Bits128 bitwiseOr(Bits128 a, Bits128 b) { ... }
public static Bits128 bitwiseXor(Bits128 a, Bits128 b) { ... }
```

```
public static Bits128 multiplyFull(long a, long b) { ... }
public static Bits128 multiplyCarryless(long a, long b) { ... }
public static Bits128 aesEncodeStep(Bits128 state, Bits128 key) { ... }
```

Suggestion: Incubate nascent new primitive and vector types similarly

package jdk.internal.vm.data.halffloat;

```
public class HalfFloat { ...
public static HalfFloat add(HalfFloat a, HalfFloat b) { ... }
public static HalfFloat sub(HalfFloat a, HalfFloat b) { ... }
public static HalfFloat mul(HalfFloat a, HalfFloat b) { ... }
public static HalfFloat sqrt(HalfFloat a) { ... }
```

public class HalfFloatVector extends AbstractVector<HalfFloat> { ... }
public class HalfFloat128Vector extends HalfFloatVector { ... }
public class HalfFloat256Vector extends HalfFloatVector { ... }
public class HalfFloat512Vector extends HalfFloatVector { ... }

Half-baked suggestion: Incubate nascent vector lane operations

package jdk.internal.vm.data;

```
public class ExtraVectorOperators extends VectorOperators { ...
   public static final Associative ADD_HALF_FLOAT = ...;
   public static final Binary AES_ENC_STEP = ...;
   public static final Binary CL_MUL_HI = ...;
   public static final Binary CL_MUL_LO = ...;
}
```

// Not clear this pans out. The point is a proving ground for ad hoc ops.

Vector atomics — what might be useful

The underlying realities of memory operations (an educated guess)

- Memory travels in cache blocks (with update masks)
- The CPU decomposes cache operations into smaller units, tracking data as scalars (words, bytes, ...)
- Existing locking operations briefly pin cache blocks and suppress decomposition
 - Sometimes they pin adjacent blocks when a locked operation crosses a line
- Presumably there are difficulties with suppressing decomposition of vectors
 - Decomposing memory operations supports value numbering and reordering
 - Suppressing decomposition breaks those optimizations
 - Decomposing operations also allows more flexible fit to internal queue limits
 - Suppressing decomposition requires more resource allocation for internal queues
- Still, it is possible, in principle, to consider cache-locked versions of vector operations

What cache-locked vector operations might look like

- Locked vector load, unmasked
 - Allocate any load queue resources, lock one or two cache lines
 - Transfer indicated data lanes into value tracking registers; release cache
- Locked vector load, masked similar to unmasked
 - · Might lock only one cache line where unmasked would lock two
- · Locked vector store, unmasked
 - Fetch unwritten parts of cache line(s), wait for all CPU data ready
 - Allocate any store queue resources, lock one or two cache lines
 - Drain store queue entries under lock; release cache
- Locked vector store, masked similar to unmasked
 - Might lock only one cache line

Existing instructions that might help



- movdqa/movdqu: Maskable; transfers up to 512 bits. Not lockable?
 - *Mostly* atomic? (Therefore *slightly* non-atomic.)
 - (Very Naive Feature Request: Allow these guys to be lock-prefixed!)
- lock cmpxchg16b: Works on *aligned* 64-bit word pairs. Slow. Improvable?
 - Could possibly help with 2-word value types.
 - Might be usable as a plain atomic load if an impossible test-value is set up.
 - Might be usable as a plain atomic store if a likely test-value is set up.
- movdir64b: Direct cache-bypassing store for 512 bits. Slow; requires fence.
 - Write-atomic, aligned. Not read-atomic; no support for reads.
 - movdiri: Like movdir64b, but 32/64 bits. Cache-bypassing, requires fence.
- TSX instructions: Rocky history. Excessive power: multiple cache lines (DCAS/MCAS).

What cache-locked vector CAS might look like

- Two vectors A, B, one mask K, one address M, returns C
 - Allocate any queue resources, lock one or two cache lines
 - Load C=(M){K}, compare to C and A under mask K
 - Store (M){K}=B under mask, but only if C{K}=A{K}
 - Drain queues; release cache
- Two vectors A, B, two masks K, K2, one address M, returns C
 - Similar to above, but store (M){K2}=B under second mask
 - Possible restriction: test mask K is limited in size/shape
- Memory transaction is always one or two-adjacent cache lines
 - Not a general DCAS/CAS2; more like a "whole cache line" CAS
 - Similar also to CMPXCHG16B, probably slow like that one.

Use cases



- Java volatile variable, with flat value >64 bits.
 - Struct-tearing is a (rare) possibility to protect against
 - Alternative to SW transactional memory (side-locks)
- Java extended VarHandle operations, on flat value >64 bits.
- Reading/writing/linking inter-thread communication or coprocessor command blocks (in user space)
- Concurrent version control on the same cache line (nR/1W SeqLock, for example).

- The alternative is object versioning: Use an indirection, CAS-patch a 64-bit pointer to current version.
- For Java volatiles, each version of the variable is a freshly allocated "buffer" object.
- One cost: Indirection harms locality (extra cache-line fetches).
- Another cost: Garbage collector has to go around cleaning up unused indirection blocks.

Plan of record



- Rely *only* on 64-bit memory atomicity.
- When in doubt, use object versioning, extra buffering, indirections.
- Invest more aggressively in one-word representations.
 - Example: value class Optional<T> { Object valOrSentinel; }
 - Instead of: value class Optional<T> { T val; boolean isPresent; }

• (Any guidance from the hardware gurus?)

Thank You

Questions? Comments?

http://cr.openjdk.java.net/~jrose/pres/ 202202-VectorTopics.pdf





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