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6.189 Multicore Programming Primer, January (IAP) 2007

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6.189 IAP 2007

Recitation 6

SIMD Programming on Cell

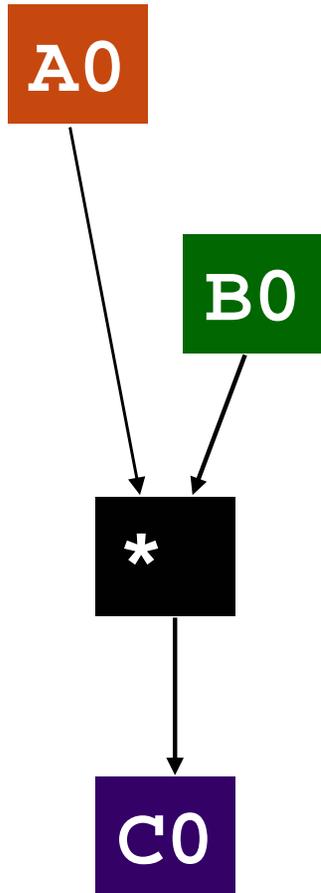
Agenda

- Overview of SIMD
- Vector Intrinsics on Cell
- SIMD Design Considerations

SIMD

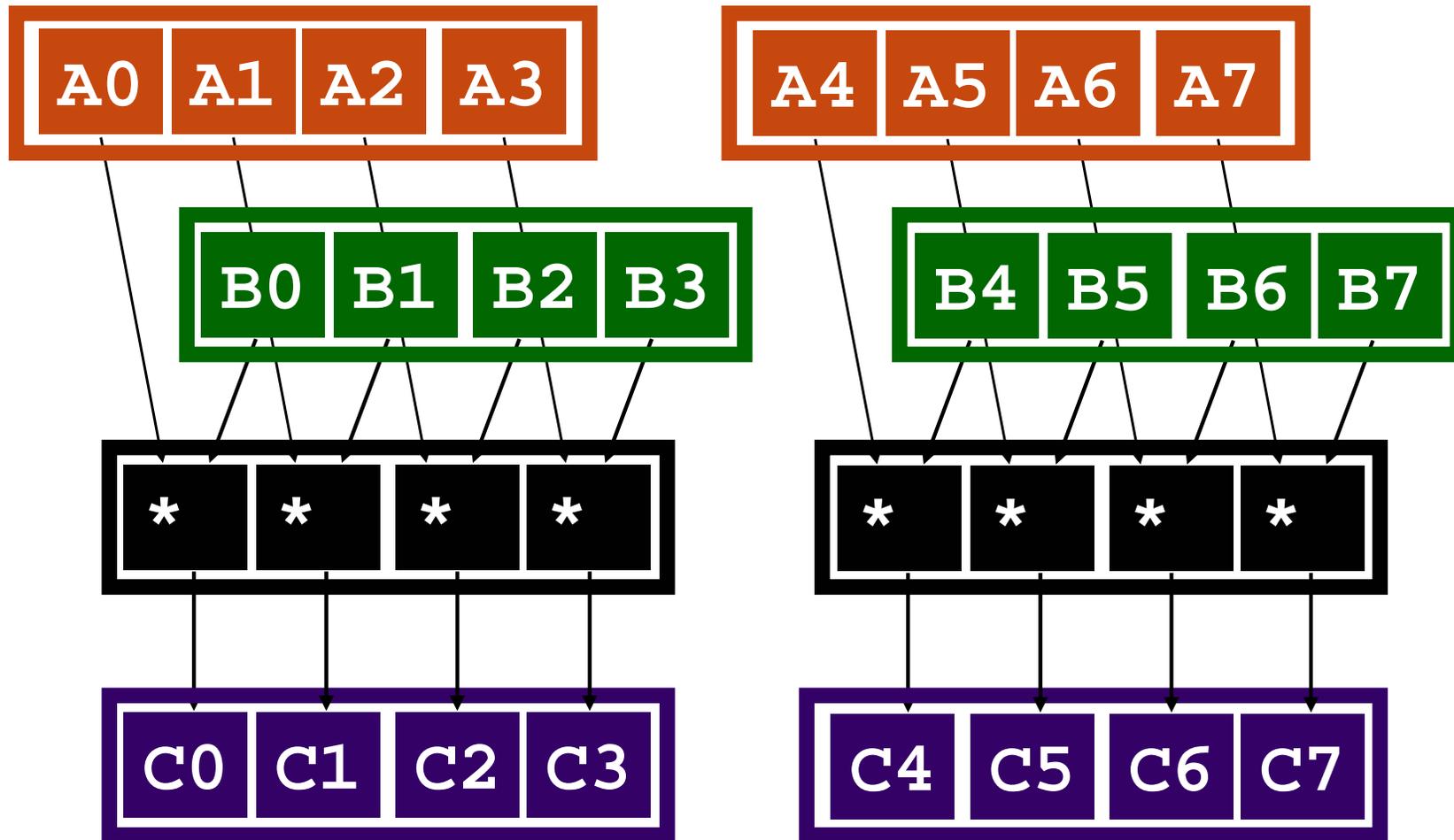
- Most compute-bound applications are performing the same computations on a lot of data
 - Dependence between iterations is rare
 - Opportunities for data parallelization across iterations and within iterations

Example: Scalar Operation



$$C[0] = A[0] * B[0]$$

Example: SIMD Vector Operation



```
for(i = 0; i < N/4; ++i)  
    C[i] = vector_mul(A[i], B[i]);
```

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Hardware Support for Data Parallelism

- Registers are 128-bits
- Can pack vectors of different data types into registers
- Operations consume and produce vector registers
 - Special assembly instructions
 - Access via C/C++ language extensions (intrinsics)

Vector Registers

- Only registers in SPU are 128-bit registers
 - Any type (including scalar types) can go into any register
- Scalar values go in a particular position in register

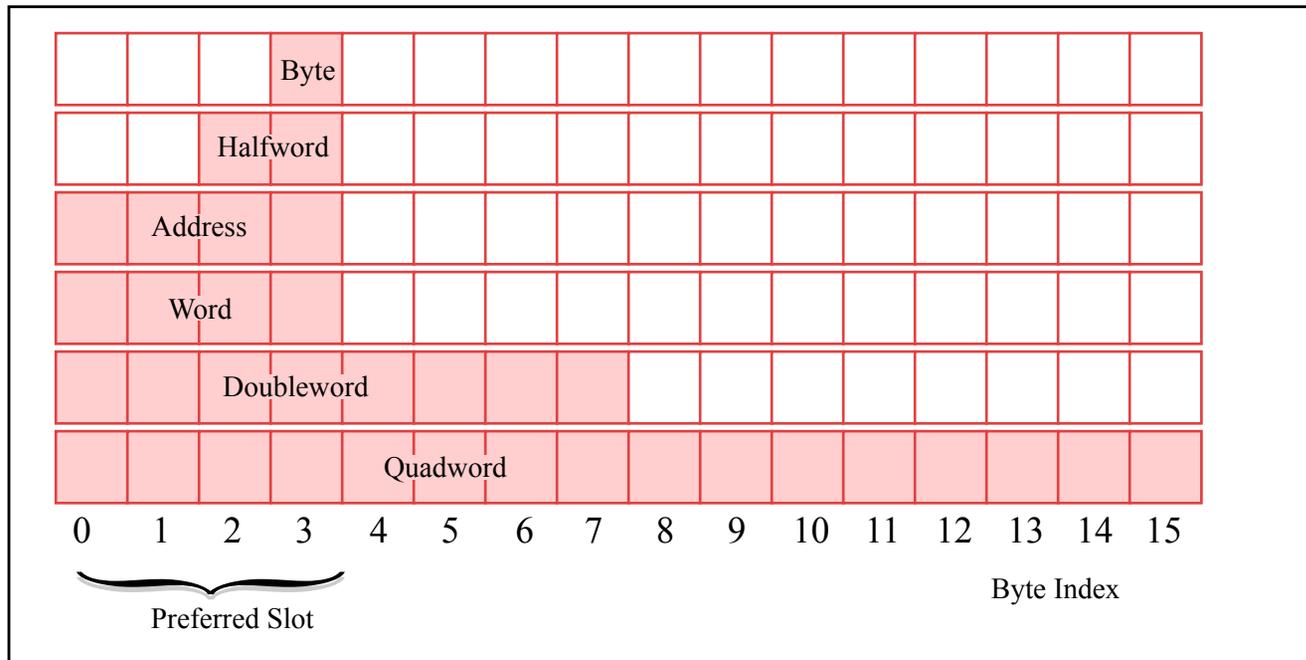


Image by MIT OpenCourseWare.

- There is overhead associated with loading and storing scalars

Writing Efficient SIMD Code

- Used the `aligned` compiler directive to control placement
 - Quadword alignment for loads and stores (`aligned(16)`)
- Transfer multiples of 16 bytes on loads and stores
 - Pad end of data if necessary

Vector Data Types

- Vector data types dictate how to interpret 128 bits
- Available on PPU and SPU:
 - 16x 8-bit int: `vector signed char`
 - 8x 16-bit int: `vector signed short`
 - 4x 32-bit int: `vector signed int`
 - 4x float: `vector float`
- Available on SPU:
 - 2x 64-bit int: `vector signed long long`
 - 2x double: `vector double`
- Pointer types, arrays, etc. work correctly

Vector Operations

- Compilers will insert vector instructions correctly for +, *, etc. when applied to vector types
- Intrinsics provide C/C++ access to vector instructions, including many which do not correspond to any operator
 - Example: `vector signed int c = spu_add(a, b);`
 - No need to worry about registers for operands
 - Looks like a function call
 - Compiler automatically generates instructions in assembly
 - Slightly different intrinsics available on PPU, SPU

Source Headers Necessary for Intrinsic

- SPU intrinsics
 - `#include <spu_intrinsics.h>`
 - `#include <spu_mfcio.h>`
- PPU intrinsics
 - `#include <ppu_intrinsics.h>`
 - `#include <vec_types.h>`

Initializing Vectors

- One of these cast notations should work (depending on your compiler):
 - `vector signed int a =
 (vector signed int)(10, 20, 30, 40);`
 - `... (vector signed int){10, 20, 30, 40};`
- Or use an intrinsic:
 - `vector signed int b = spu_splats(20);
 // Same as (20, 20, 20, 20)`

Accessing Vector Elements

- `typedef union {
 int v[4];
 vector signed int vec;
} intVec;`
- Unpack scalars from vector:
 - `intVec a;
a.vec = ...;
... = a.v[2];`
 - `... = spu_extract(va, 2);`
- Pack scalars into vector:
 - `a.v[0] = ...; a.v[1] = ...;
a.v[2] = ...; a.v[3] = ...;
... = a.vec;`

Interpret a segment of memory either as an array...

`v[0]` `v[1]` `v[2]` `v[3]`

or as a vector type...

`vec`

so that values written in one format can be read in the other

Vector Operations

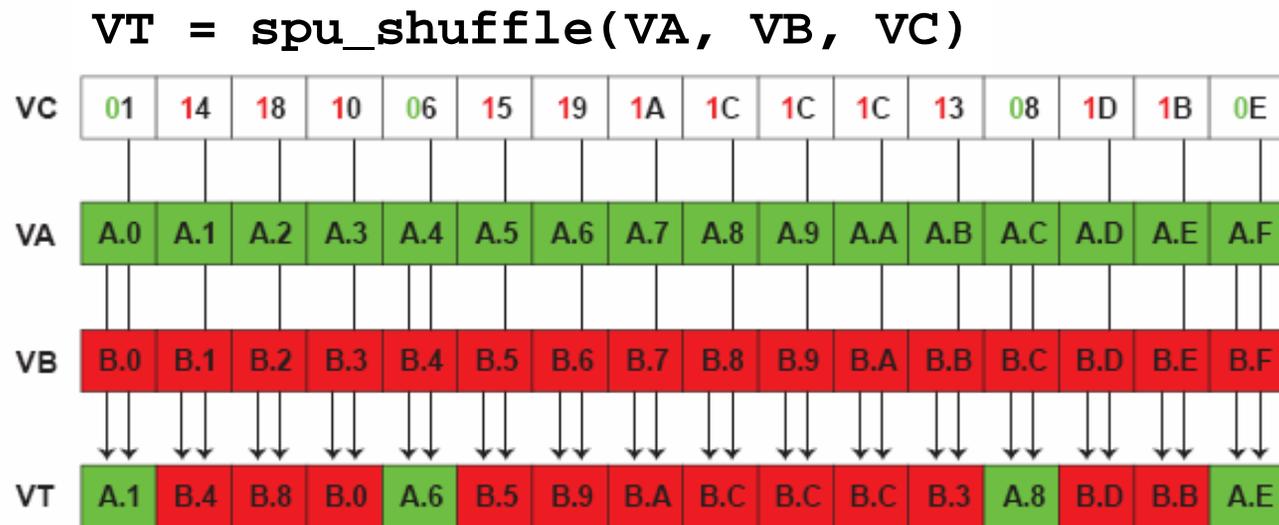
- Integer instructions
 - Floating-point instructions
 - Permutation/formatting instructions
 - Load and store instructions
-
- Complete reference available from course web site

Vector Arithmetic and Logical Operations

- PPU
 - `vec_add`, `vec_sub`, `vec_madd`, ...
 - `vec_and`, `vec_or`, `vec_xor`, ...
- SPU
 - `spu_add`, `spu_sub`, `spu_madd`,
`spu_mul`, `spu_re`, ...
 - `spu_and`, `spu_or`, `spu_xor`, ...
- Integer/FP operation associated with the correct vector types (char, int, float, etc.) is usually automatically selected by the compiler

Vector Shuffle Operation

- Rearrange bytes of vectors: `spu_shuffle(A, B, pattern)`
 - Each byte of the output is one of the bytes of **A** or **B**
 - For each byte of output, corresponding byte of pattern specifies which byte of **A** or **B** to copy
 - Bit 4 of each pattern byte specifies **A** or **B**
 - Bits 0-3 (4 low-order bits) of each pattern byte specify which byte (0-15) of source to take
 - Ex: 2nd byte of pattern is 0x14, so take byte 4 from **B**



Vector Shuffle Operation

- Generating the shuffle pattern:

```
pattern =  
    (vector unsigned char)( b0, b1, b2, b3,  
                            b4, b5, b6, b7,  
                            b8, b9, b10, b11,  
                            b12, b13, b14, b15);
```

- Example: reverse the order of bytes in a

```
a = spu_shuffle(a, a,  
                (vector unsigned char)(15, 14, 13, 12,  
                                        11, 10, 9, 8,  
                                        7, 6, 5, 4,  
                                        3, 2, 1, 0));
```

Vector Rotate Operations

- Rotate shifts vector elements left or right
 - `spu_rl(v, count)`
 - `vec_rl(v, count)`

Review: *sim* (Recitation 2)

- Simple 3D gravitational body simulator
- n objects, each with mass, initial position, initial velocity

```
float mass[NUM_BODIES];  
VEC3D pos[NUM_BODIES];  
VEC3D vel[NUM_BODIES];
```

```
typedef struct _VEC3D {  
    float x, y, z;  
} VEC3D;
```

- Simulate motion using Euler integration
 - Calculate the force of each object on every other
 - Calculate net force on and acceleration of each object
 - Update position

```
VEC3D d;  
// Calculate displacement from i to j  
d.x = pos[j].x - pos[i].x;  
d.y = pos[j].y - pos[i].y;  
d.z = pos[j].z - pos[i].z;
```

Re-engineering for SIMD

- One approach to SIMD: array of structs
 - Pad each (x, y, z) vector to fill a quadword
 - Components (x, y, z) correspond to first three words of a vector float
 - Quadwords for different vectors stored consecutively

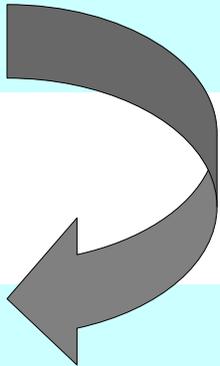


```
typedef union _VEC3D {  
    struct {float x, y, z;};  
    vector float vec;  
} QWORD_ALIGNED VEC3D;
```

Re-engineering for SIMD

- Now we can replace component-wise addition, subtraction, and multiplication with SIMD instructions

```
VEC3D d;  
// Calculate displacement from i to j  
d.x = pos[j].x - pos[i].x;  
d.y = pos[j].y - pos[i].y;  
d.z = pos[j].z - pos[i].z;
```



```
vector float d;  
// Calculate displacement from i to j  
d = spu_sub(pos[j].vec, pos[i].vec);
```

Exercise 1 (15 minutes)

- Complete the SIMD implementation of *sim*
 - See example code in recitations section.
 - `tar xzf rec6.tar.gz`
 - `cd rec6/sim_aos`
 - `export CELL_TOP=/opt/ibm/cell-sdk/prototype`
- `spu/sim_spu.c`, line 49: implement `eltsumf4()`
 - Given a vector float (a, b, c, d) , return the vector float $(a+b+c+d, a+b+c+d, a+b+c+d, a+b+c+d)$
 - You can do this with two shuffles and two adds
 - Note `vec_float4` is shorthand for `vector float`
 - Check your results with `./sim 1`
 - Will print “Verify succeeded” if your implementation is correct

Exercise 1

- Solution is in `sim_aos_soln`
- Sample implementation:

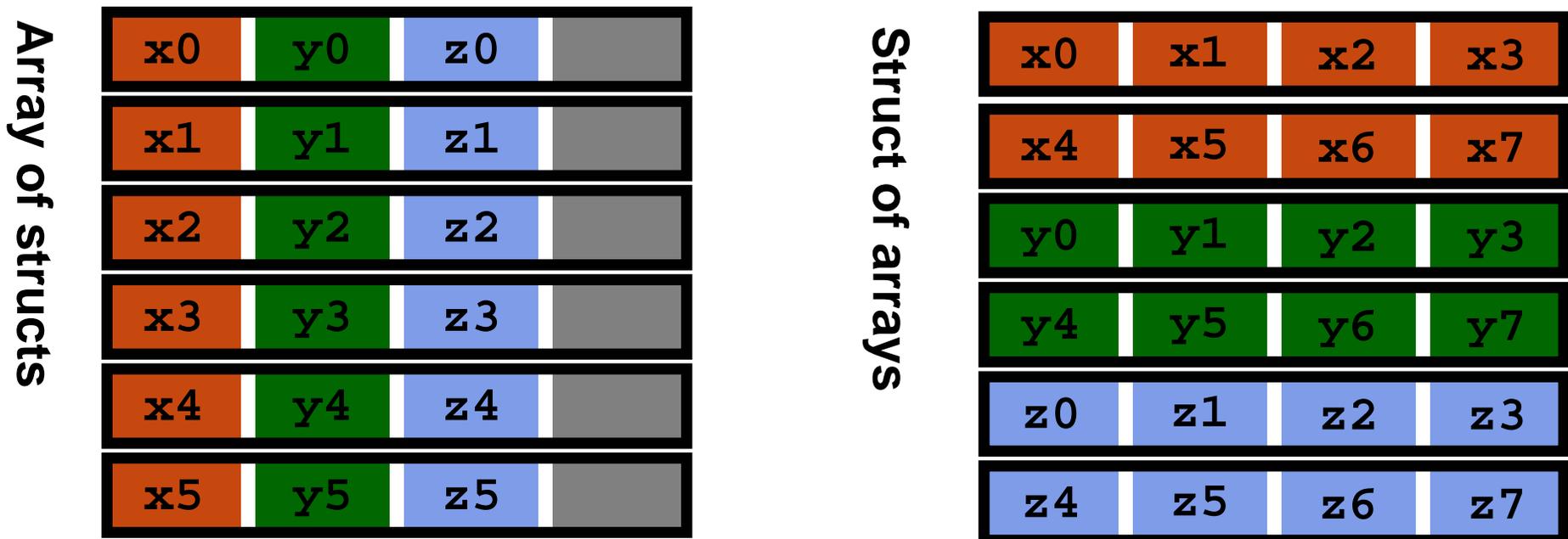
```
vec_float4 b;  
  
b = spu_shuffle(a, a,  
               (vector unsigned char)(4, 5, 6, 7, 0, 1, 2, 3,  
                                       12, 13, 14, 15, 8, 9, 10, 11));  
a = spu_add(a, b);  
  
b = spu_shuffle(a, a,  
               (vector unsigned char)(8, 9, 10, 11, 12, 13, 14, 15,  
                                       0, 1, 2, 3, 4, 5, 6, 7));  
a = spu_add(a, b);  
  
return a;
```

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- **SIMD Design Considerations**

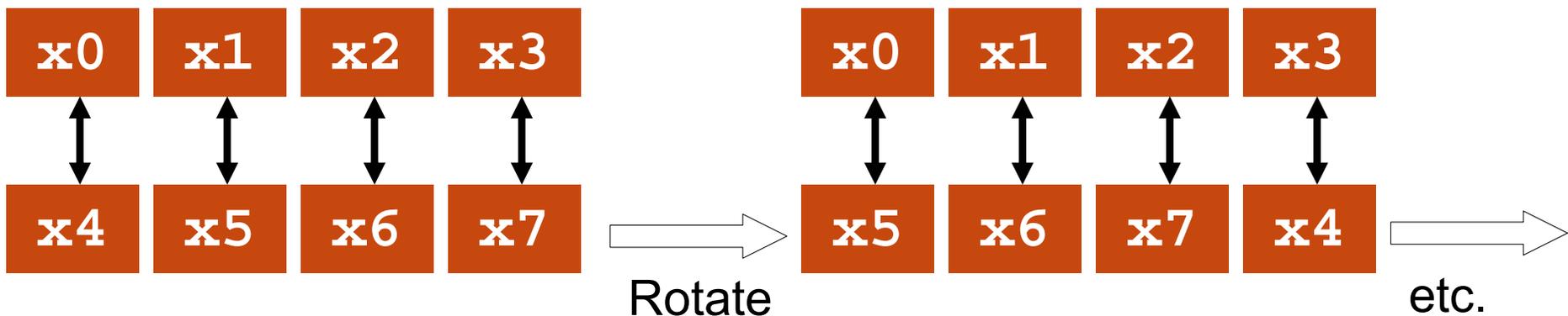
SIMD Design Considerations

- Data layout: struct of arrays vs. array of structs
 - Exercise 1 used an AOS layout
 - Alternatively we could use a SOA layout to lay out fields consecutively
 - Can apply different algorithms on new data layout



Struct of Array Layout

- Need 12 quadwords to store state for 8 objects
 - x , y , z position and velocity components
 - No padding component needed in SOA
- For each component, do four pair-interactions at once with SIMD instructions
 - Rotate quadword 3 more times to get all 16 pair-interactions between two quadwords



Performance Results Summary

- Example code in `rec6/sim_soa`
- 6144 objects, compiled with `-O2`
- Time per simulation step
 - SIMD array of structs: 300 ms
 - SIMD struct of arrays: 80 ms

Summary of Cell Optimizations That Were Covered

- Baseline native code was sequential and scalar
 - Scalar (PPU): 1510 ms (`rec6/sim_spu, -O3`)
- Parallelized code with double buffering for SPUs
 - Scalar (6 SPUs): 420 ms (`rec6/sim_db`)
- Applied SIMD optimizations
 - SIMD array of structs: 300 ms (`rec6/sim_aos_soln`)
- Redesigned algorithm to better suite SIMD parallelism
 - SIMD struct of arrays: 80 ms (`rec6/sim_soa`)
- Overall speedup compared to native sequential execution
 - Expected: ~24x (6 SPUs * 4 way SIMD)
 - Achieved: 18x