

Recitation 6

SIMD Programming on Cell

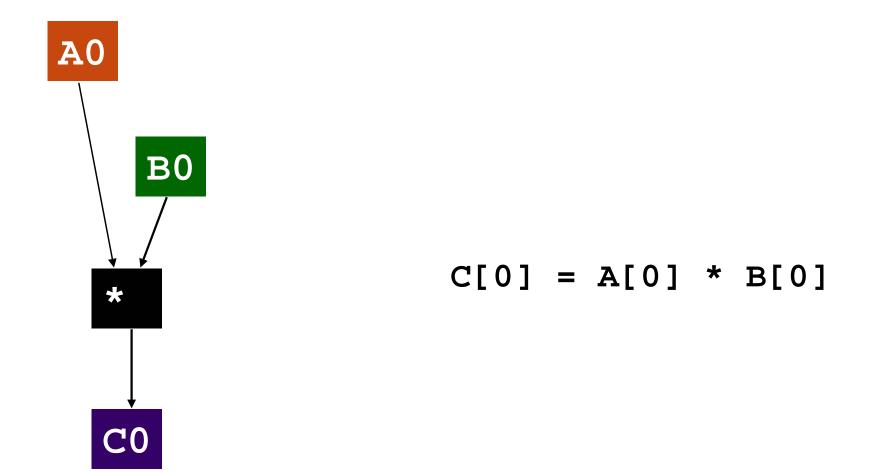
Agenda

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

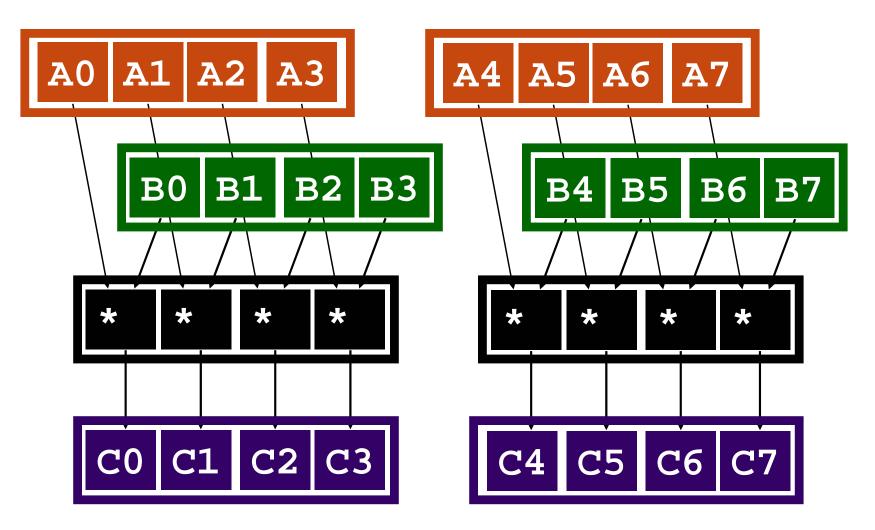


- Most compute-bound applications are performing the same computations on a lot of data
 - Dependence between iterations is rare
 - Opportunities for data parallelization across iterations and within iterations

Example: Scalar Operation



Example: SIMD Vector Operation



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



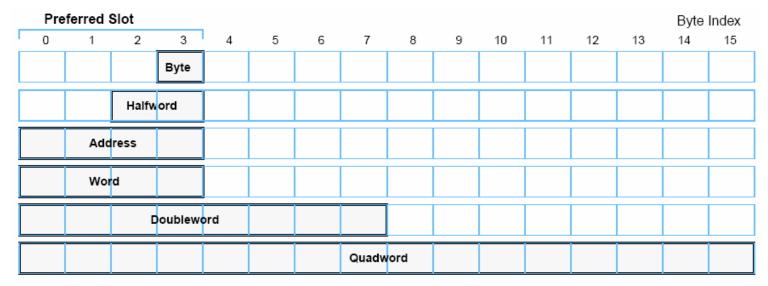
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Hardware Support for Data Parallelism

- 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



 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

- 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 Intrinsics

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

Initializing Vectors

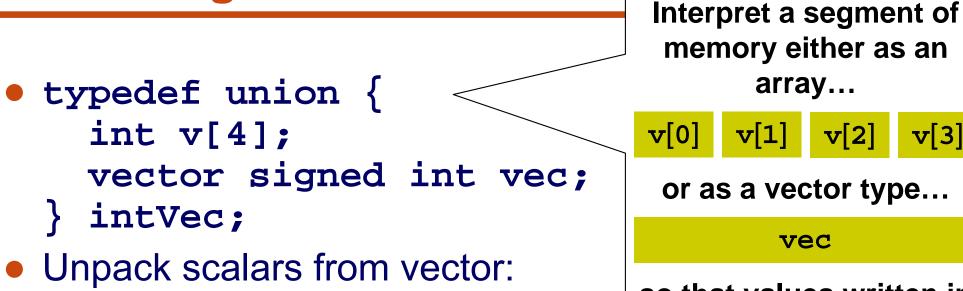
- One of these cast notations should work (depending on your compiler):
 - vector signed int a =

(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



intVec a;

```
a.vec = ...;
```

$$= a.v[2];$$

... = spu_extract(va, 2);

Pack scalars into vector:

or as a vector type...

v[3]

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

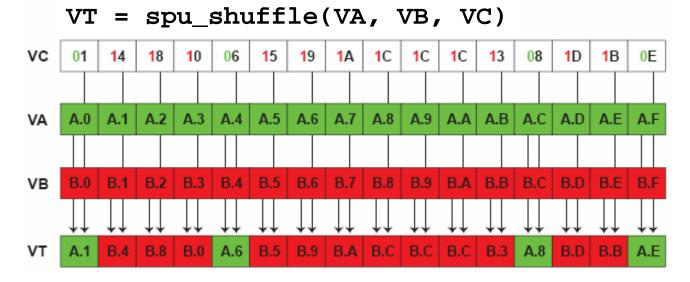
- vec_add, vec_sub, vec_madd, ...
- vec_and, vec_or, vec_xor, ...

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



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Vector Shuffle Operation

• Generating the shuffle pattern:

Example: reverse the order of bytes in a

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];

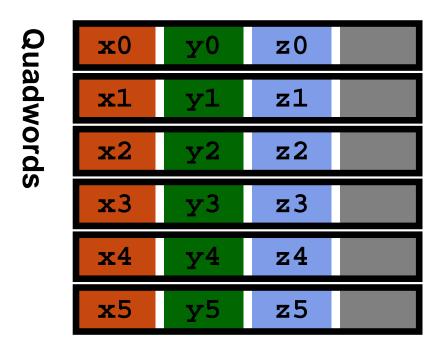
typedef struct _VEC3D {
 float x, y, z;
} VEC3D;

- 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

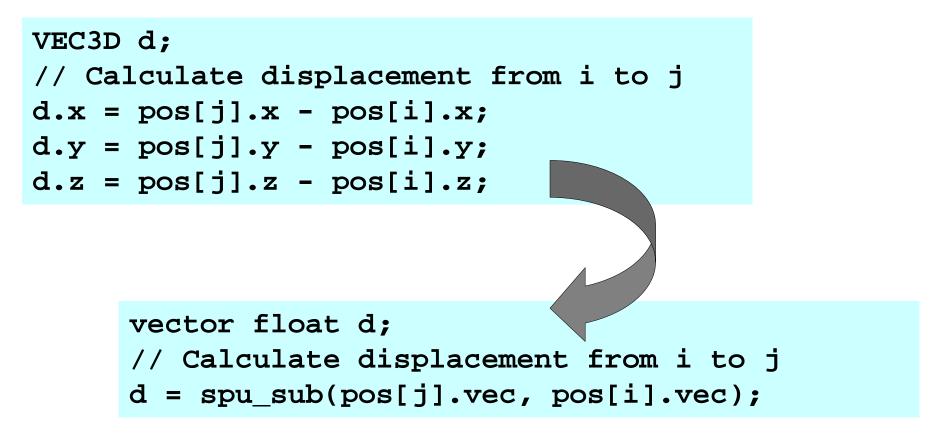
- One approach to SIMD: array of structs
 - Pad each (x, y, z) vector to fill a quadword
 - Components (x, y, z) correspond to first three words of a vector float
 - Quadwords for different vectors stored consecutively



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



Exercise 1 (15 minutes)

- Complete the SIMD implementation of *sim*
 - wget http://cag.csail.mit.edu/ps3/recitation6/rec6.tar.gz
 - 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:

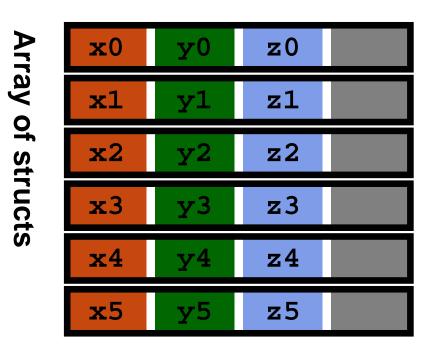


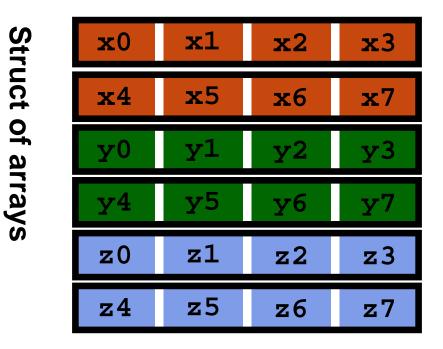
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SIMD Design Considerations

• 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

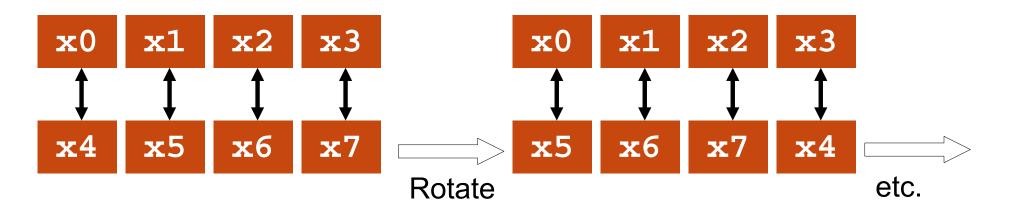




Struct of Array Layout

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 –02
- 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): 1510ms (rec6/sim_spu, -03)
- 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

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