CS425
Computer Systems Architecture

Fall 2021

Vector Processors

Flynn's Taxonomy of Computers

- Mike Flynn, "Very High-Speed Computing Systems," Proc. of IEEE, 1966

- SISD: Single instruction operates on single data element
- SIMD: Single instruction operates on multiple data elements
  - Vector processor
- MISD: Multiple instructions operate on single data element
  - Closest form: systolic array processor, streaming processor
- MIMD: Multiple instructions operate on multiple data elements (multiple instruction streams)
  - Multiprocessor
  - Multithreaded processor

Data Parallelism

- Concurrency arises from performing the same operation on different pieces of data
  - Single instruction multiple data (SIMD)
  - E.g., dot product of two vectors
- Contrast with data flow
  - Concurrency arises from executing different operations in parallel (in a data driven manner)
- Contrast with thread (“control”) parallelism
  - Concurrency arises from executing different threads of control in parallel
- SIMD exploits operation-level parallelism on different data
  - Same operation concurrently applied to different pieces of data
  - A form of ILP where instruction happens to be the same across data

Vector Processors (1/2)

- A vector is a one-dimensional array of numbers
- Many scientific/commercial programs use vectors

```
for (i = 0; i<=49; i++)
  C[i] = (A[i] + B[i]) / 2
```

- A vector processor is one whose instructions operate on vectors rather than scalar (single data) values
- Basic requirements
  - Need to load/store vectors → vector registers (contain vectors)
  - Need to operate on vectors of different lengths → vector length register (VLEN)
  - Elements of a vector might be stored apart from each other in memory → vector stride register (VSTR)
    - Stride: distance in memory between two elements of a vector

Vector Processors (1/2)

- A vector instruction performs an operation on each element in consecutive cycles
  - Vector functional units are pipelined
  - Each pipeline stage operates on a different data element

- Vector instructions allow deeper pipelines
  - No intra-vector dependencies → no hardware interlocking needed within a vector
  - No control flow within a vector
  - Known stride allows easy address calculation for all vector elements
    - Enables prefetching of vectors into registers/cache/memory

Vector Processor Properties

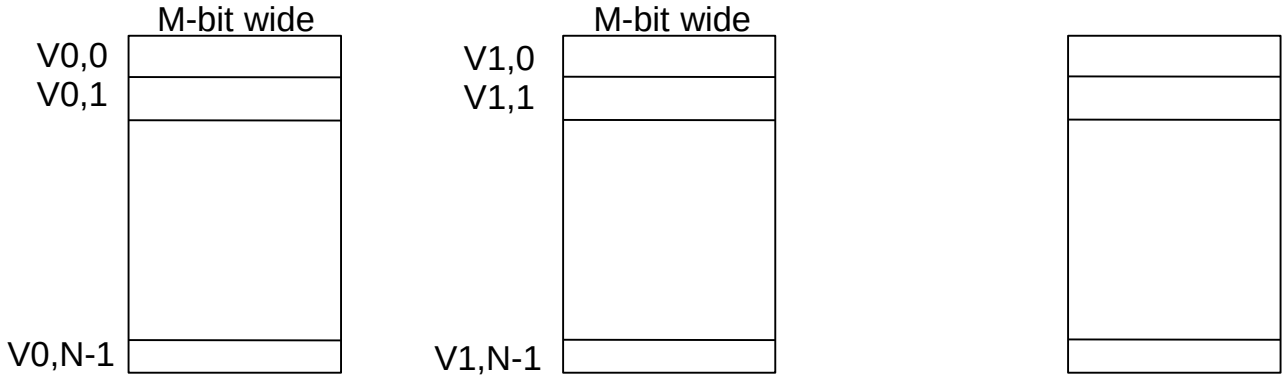
- No dependencies within a vector
  - Pipelining & parallelization work really well
  - Can have very deep pipelines, no dependencies!
- Each instruction generates a lot of work
  - Reduces instruction fetch bandwidth requirements
- Highly regular memory access pattern
- No need to explicitly code loops
  - Fewer branches in the instruction sequence
- Works (only) if parallelism is regular (data/SIMD parallelism)
  - Many vector operations
  - Very inefficient if parallelism is irregular

Vector Processor Limitations

- Memory (bandwidth) can easily become a bottleneck, especially if
  - compute/memory operation balance is not maintained
  - data is not mapped appropriately to memory banks

Vector Registers

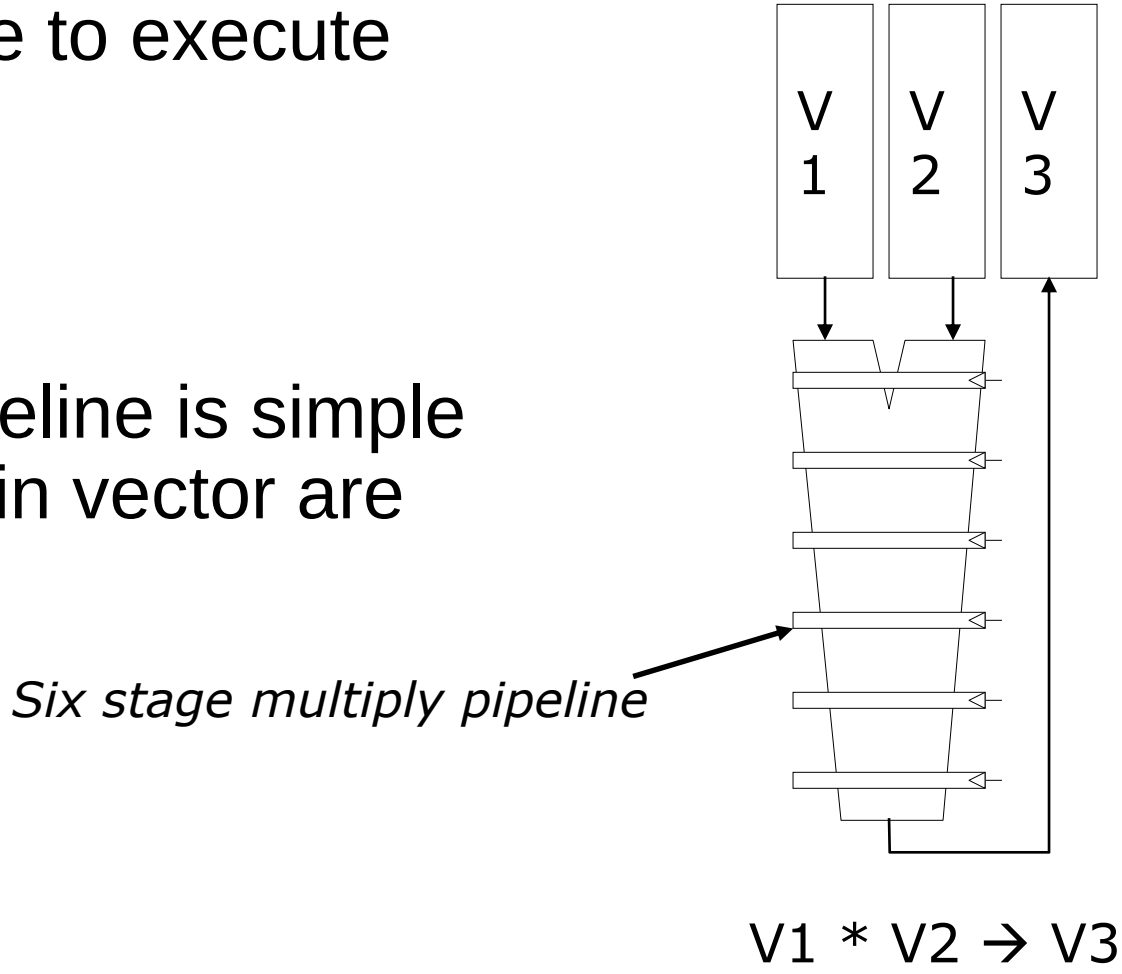
- Each vector data register holds N M-bit values
- Vector control registers: VLEN, VSTR, VMASK
- Maximum VLEN can be N
  - Maximum number of elements stored in a vector register
- Vector Mask Register (VMASK)
  - Indicates which elements of vector to operate on
  - Set by vector test instructions
    - e.g., VMASK[i] = (V_k[i] == 0)



Vector Functional Units

- Use a deep pipeline to execute element operations
  → fast clock cycle

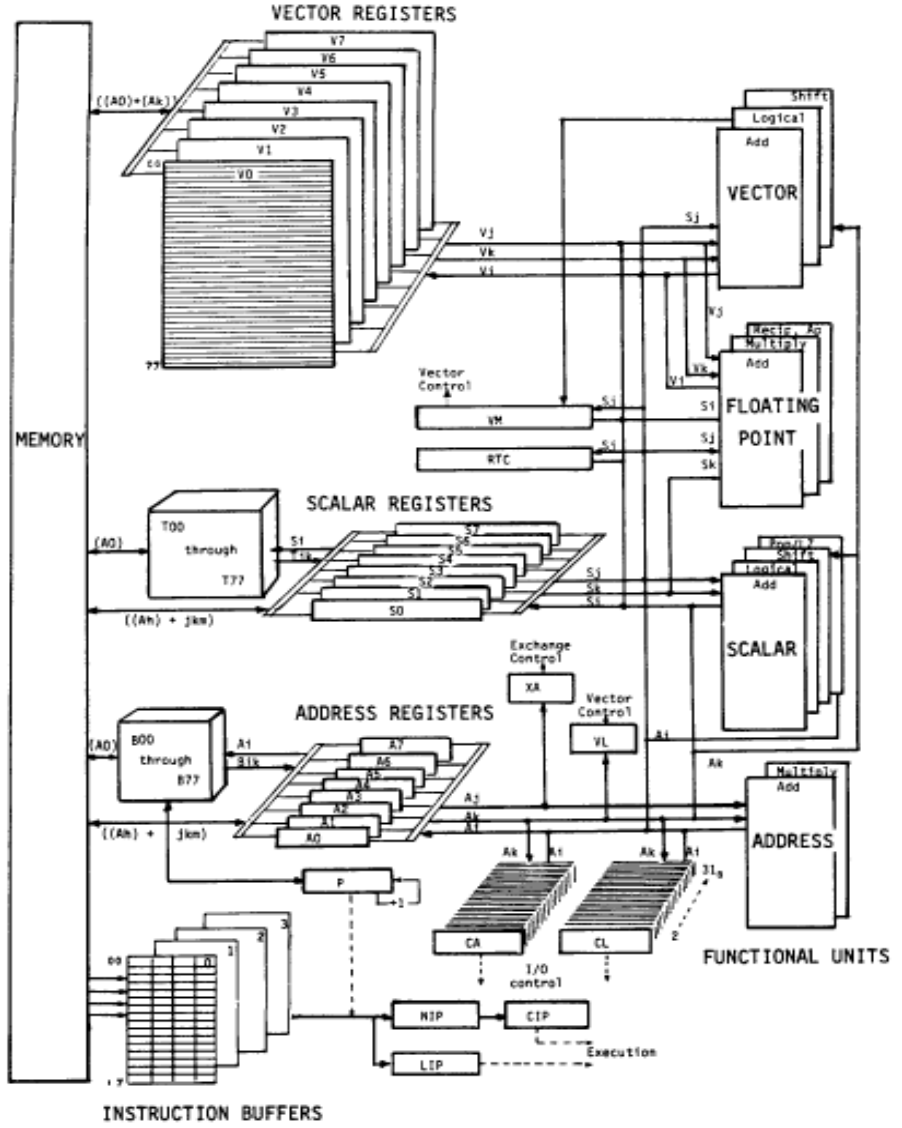
- Control of deep pipeline is simple because elements in vector are independent



Vector Machine Organization (CRAY-1)

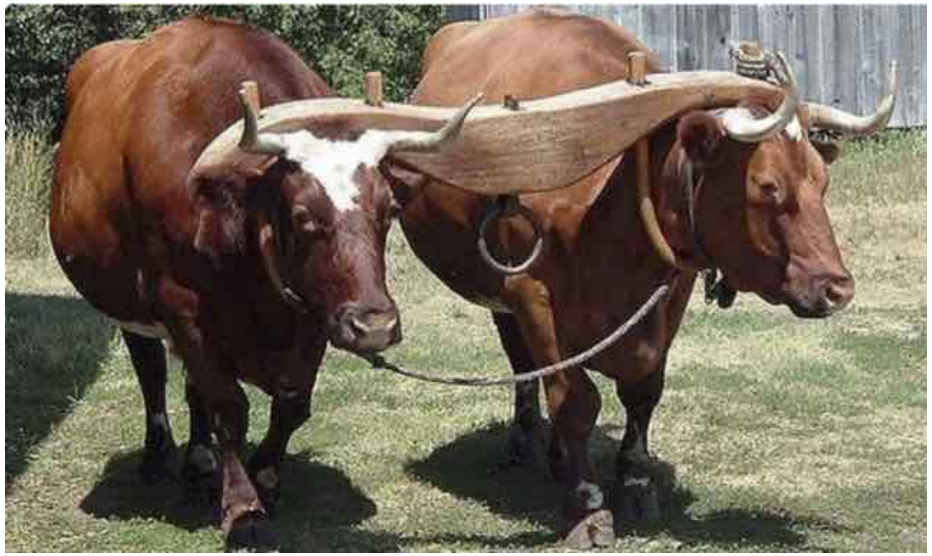
- CRAY-1
- Russell, "The CRAY-1 computer system," CACM 1978.

- Scalar and vector modes
- 8 64-element vector registers
- 64-bits per element
- 16 memory banks
- 8 64-bit scalar registers
- 8 24-bit address registers



Seymour Cray, the Father of Supercomputers

"If you were plowing a field, which would you rather use: Two strong oxen or 1024 chickens?"

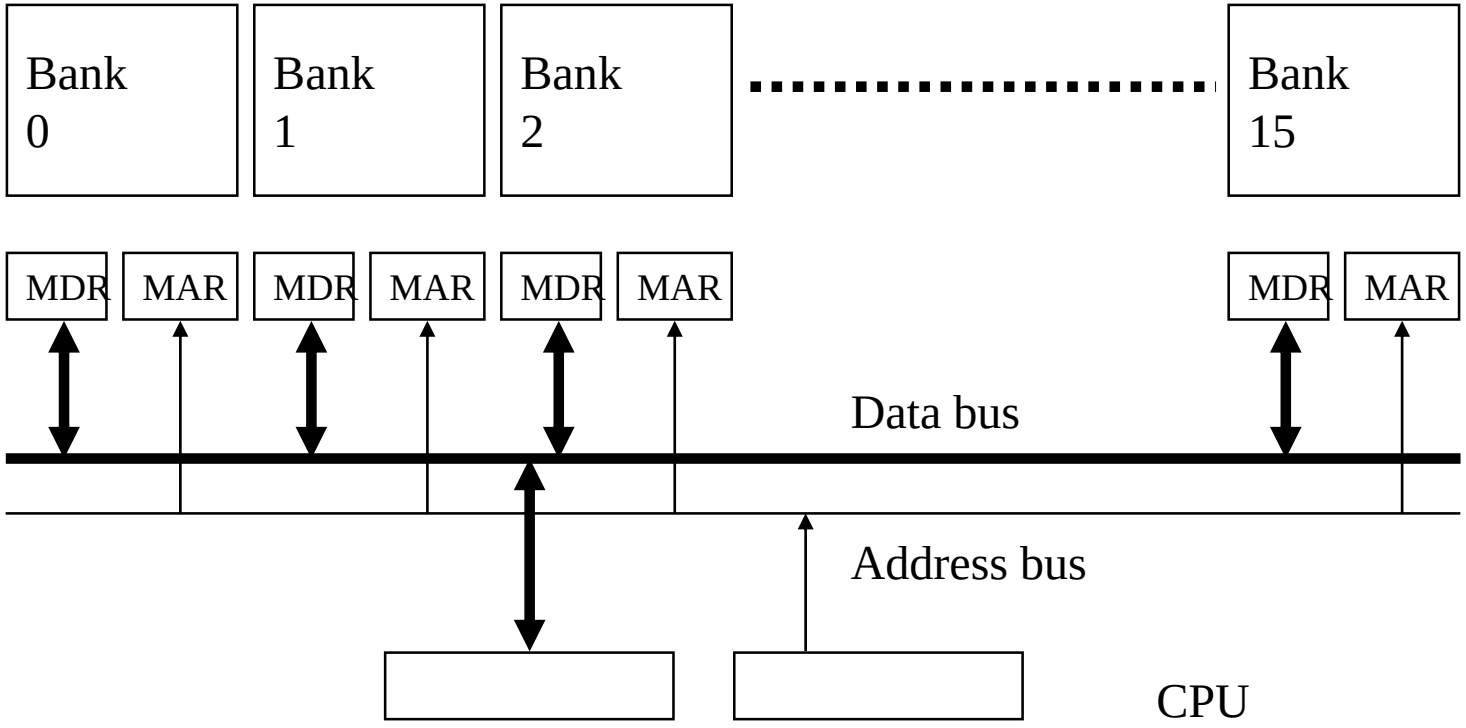


Loading/Storing Vectors from/to Memory

- Requires loading/storing multiple elements
- Elements separated from each other by a constant distance (stride)
  - Assume stride = 1 for now
- Elements can be loaded in consecutive cycles if we can start the load of one element per cycle
  - Can sustain a throughput of one element per cycle
- Question: How do we achieve this with a memory that takes more than 1 cycle to access?
- Answer: Bank the memory; interleave the elements across banks

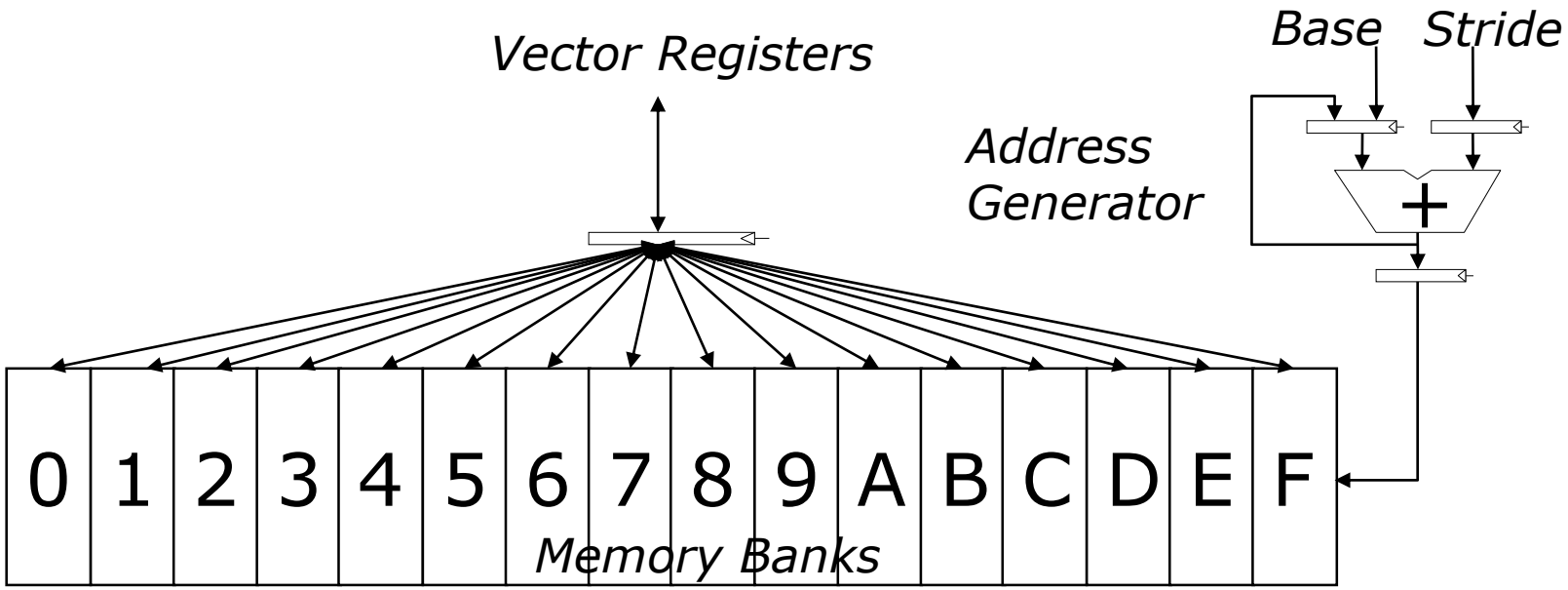
Memory Banking

- Memory is divided into banks that can be accessed independently; banks share address and data buses (to minimize pin cost)
- Can start and complete one bank access per cycle
- Can sustain N parallel accesses if all N go to different banks



Vector Memory System

- Next address = Previous address + Stride
- If (stride == 1) && (consecutive elements interleaved across banks) && (number of banks >= bank latency), then
  - we can sustain 1 element/cycle throughput



Scalar Code Example: Element-Wise Avg.

- For i = 0 to 49
  - C[i] = (A[i] + B[i]) / 2

- Scalar code (instruction and its latency)

```
        MOVI R0 = 50                1
        MOVA R1 = A                 1
        MOVA R2 = B                 1
        MOVA R3 = C                 1
   X:   LD R4 = MEM[R1++]           11   //autoincrement addressing
        LD R5 = MEM[R2++]           11
        ADD R6 = R4 + R5            4
        SHFR R7 = R6 >> 1           1
        ST MEM[R3++] = R7           11
        DECBNZ R0, X                 2   //decrement and branch if NZ
```

304 dynamic instructions

Scalar Code Execution Time (In Order)

- Scalar execution time on an in-order processor with 1 bank
  - First two loads in the loop cannot be pipelined: 2*11 cycles
  - 4 + 50*40 = 2004 cycles

- Scalar execution time on an in-order processor with 16 banks (word-interleaved: consecutive words are stored in consecutive banks)
  - First two loads in the loop can be pipelined
  - 4 + 50*30 = 1504 cycles

- Why 16 banks?
  - 11-cycle memory access latency
  - Having 16 (>11) banks ensures there are enough banks to overlap enough memory operations to cover memory latency

Vectorizable Loops

- A loop is vectorizable if each iteration is independent of any other

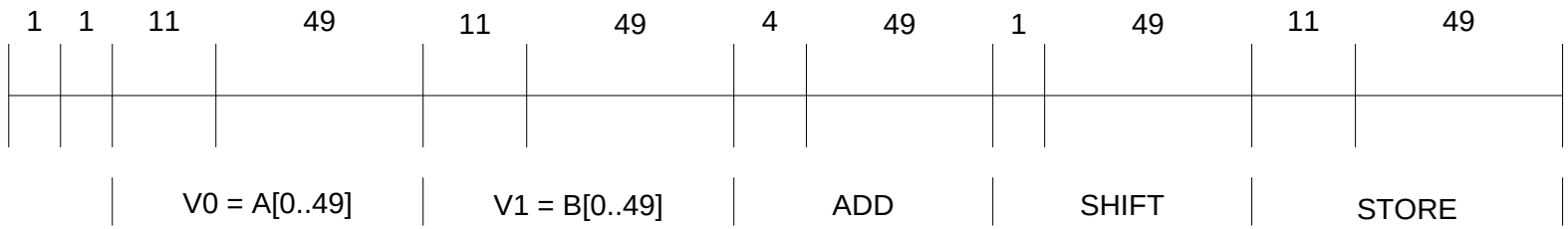
- For i = 0 to 49
  - C[i] = (A[i] + B[i]) / 2
- Vectorized loop (each instruction and its latency):

MOVI VLEN = 50	1
MOVI VSTR = 1	1
VLD V0 = A	11 + VLEN – 1
VLD V1 = B	11 + VLEN – 1
VADD V2 = V0 + V1	4 + VLEN – 1
VSHFR V3 = V2 >> 1	1 + VLEN – 1
VST C = V3	11 + VLEN – 1

7 dynamic instructions

Basic Vector Code Performance

- Assume no chaining (no vector data forwarding)
  - i.e., output of a vector functional unit cannot be used as the direct input of another
  - The entire vector register needs to be ready before any element of it can be used as part of another operation
- One memory port (one address generator)
- 16 memory banks (word-interleaved)

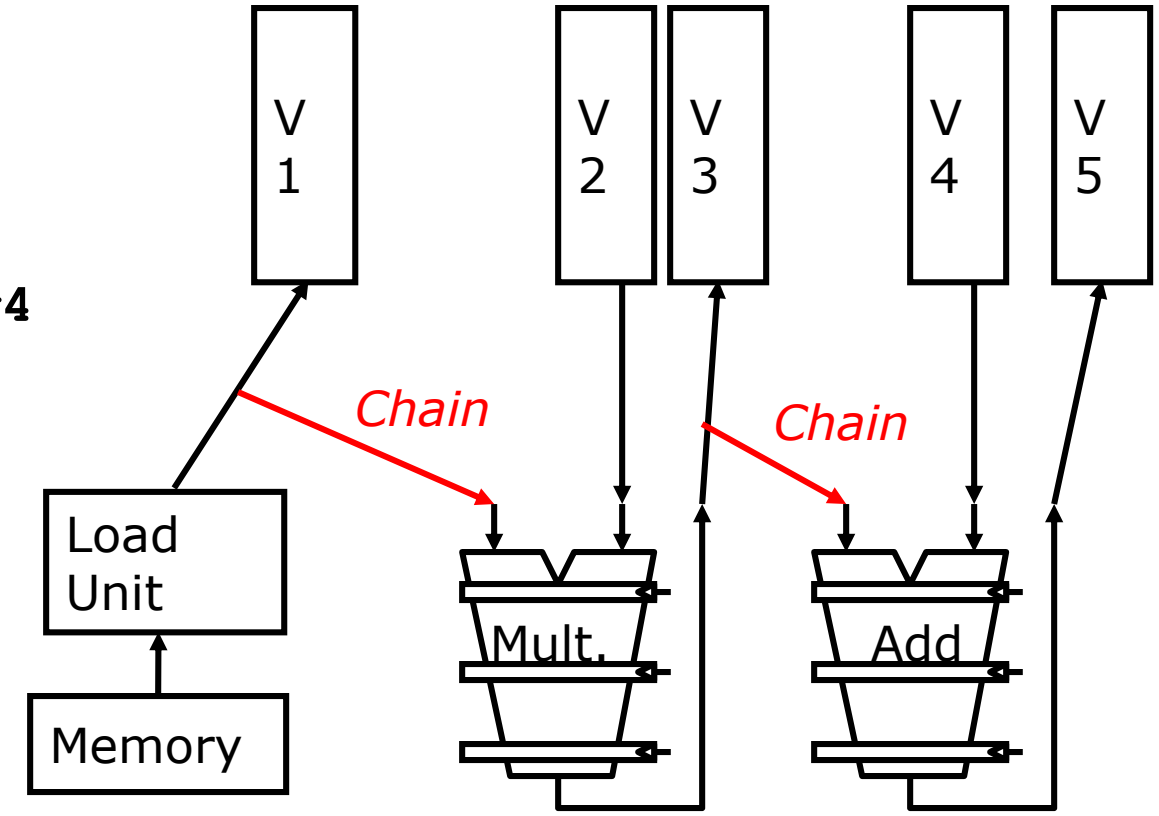


- 285 cycles

Vector Chaining

- Vector chaining: Data forwarding from one vector functional unit to another

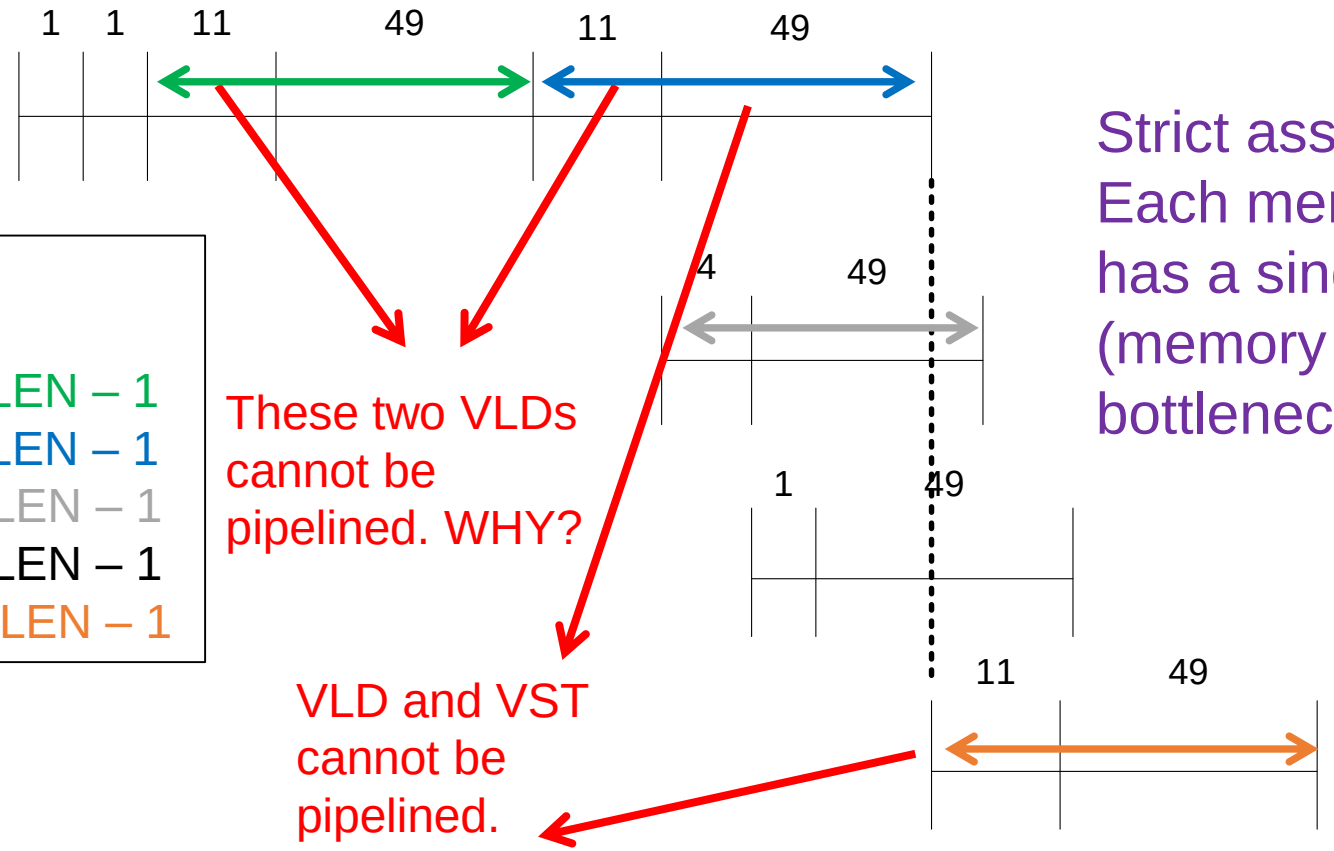
```
LV   v1
MULV v3,v1,v2
ADDV v5, v3, v4
```



Vector Code Performance - Chaining

- Vector chaining: Data forwarding from one vector functional unit to another

MOVI VLEN = 50	1
MOVI VSTR = 1	1
VLD V0 = A	11 + VLEN – 1
VLD V1 = B	11 + VLEN – 1
VADD V2 = V0 + V1	4 + VLEN – 1
VSHFR V3 = V2 >> 1	1 + VLEN – 1
VST C = V3	11 + VLEN – 1



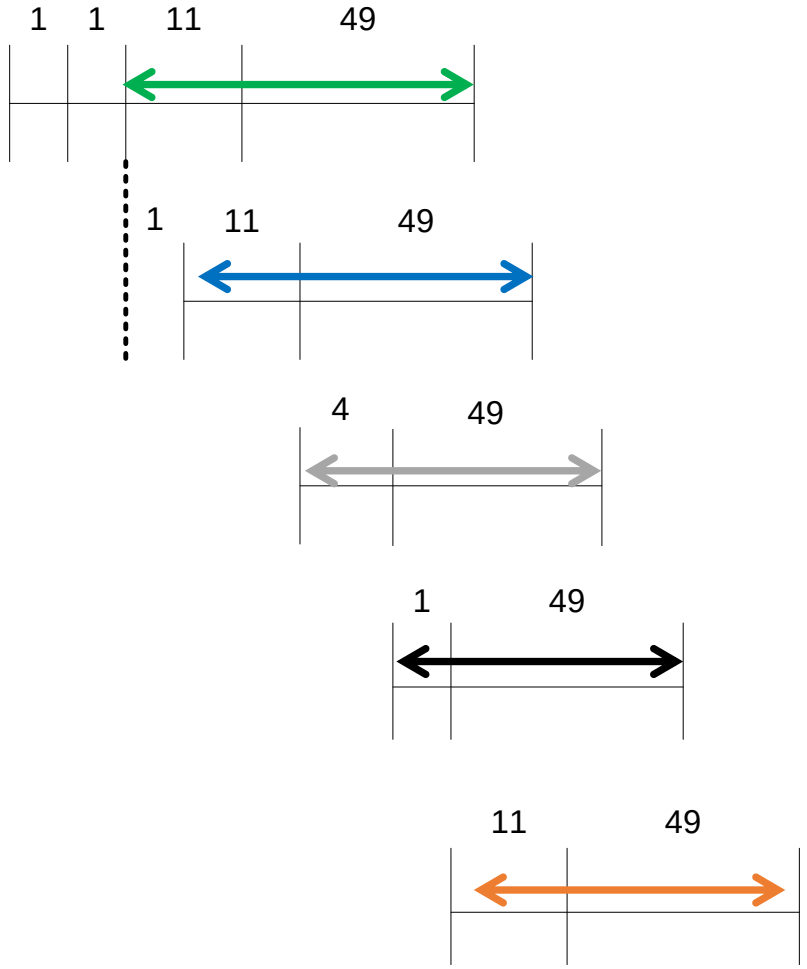
Strict assumption: Each memory bank has a single port (memory bandwidth bottleneck)

- 182 cycles

Vector Code Performance – Multiple Memory Ports

- Chaining and 2 load ports, 1 store port in each bank

MOVI VLEN = 50	1
MOVI VSTR = 1	1
VLD V0 = A	11 + VLEN – 1
VLD V1 = B	11 + VLEN – 1
VADD V2 = V0 + V1	4 + VLEN – 1
VSHFR V3 = V2 >> 1	1 + VLEN – 1
VST C = V3	11 + VLEN – 1



- 79 cycles
- 19x perf. improvement!
  - *was 1504 cycles for scalar*

Data Element Num vs. Max Vector Length

- What if # data elements > # elements in a vector register?
  - Idea: Break loops so that each iteration operates on # elements in a vector register
    - E.g., 527 data elements, 64-element VREGs
    - 8 iterations where VLEN = 64
    - 1 iteration where VLEN = 15 (need to change value of VLEN)
  - Called vector stripmining

Irregular Data Layout?

- What if vector data is not stored in a strided fashion in memory? (irregular memory access to a vector)
  - Idea: Use indirection to combine/pack elements into vector registers
  - Called scatter/gather operations

Gather/Scatter Operations

Want to vectorize loops with indirect accesses:

```
for (i=0; i<N; i++)
    A[i] = B[i] + C[D[i]]
```

Indexed load instruction (*Gather*)

```
LV vD, rD         # Load indices in D vector
LVI vC, rC, vD    # Load indirect from rC base
LV vB, rB         # Load B vector
ADDV.D vA,vB,vC   # Do add
SV vA, rA         # Store result
```

Gather/Scatter Operations

- Gather/scatter operations often implemented in hardware to handle sparse vectors (matrices)
- Vector loads and stores use an index vector which is added to the base register to generate the addresses
- *Scatter* example

Index Vector	Data Vector (to Store)
0	3.14
2	6.50
6	71.20
7	2.71

Stored Vector (in Memory)	
Base+0	3.14
Base+1	X
Base+2	6.50
Base+3	X
Base+4	X
Base+5	X
Base+6	71.20
Base+7	2.71

Conditional Operations in a Loop

- What if some operations should not be executed on a vector (based on a dynamically-determined condition)?

```
loop:       for (i=0; i<N; i++)
                    if (a[i] != 0) then b[i]=a[i]*b[i]
```

- Idea: Masked operations
  - VMASK register is a bit mask determining which data element should not be acted upon

```
VLD V0 = A
VLD V1 = B
VMASK = (V0 != 0)
VMUL V1 = V0 * V1
VST B = V1
```

  - This is predicated execution. Execution is *predicated* on mask bit.

Another Example with Masking

```
for (i = 0; i < 64; ++i)
    if (a[i] >= b[i])
        c[i] = a[i]
    else
        c[i] = b[i]
```

A	B	VMASK
1	2	0
2	2	1
3	2	1
4	10	0
-5	-4	0
0	-3	1
6	5	1
-7	-8	1

Steps to execute the loop in SIMD code

1. Compare A, B to get VMASK

2. Masked store of A into C

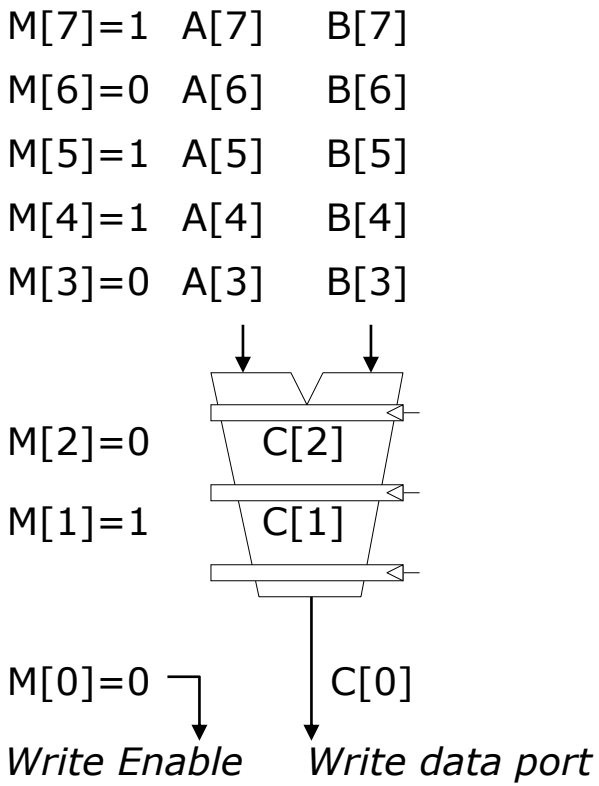
3. Complement VMASK

4. Masked store of B into C

Masked Vector Instructions

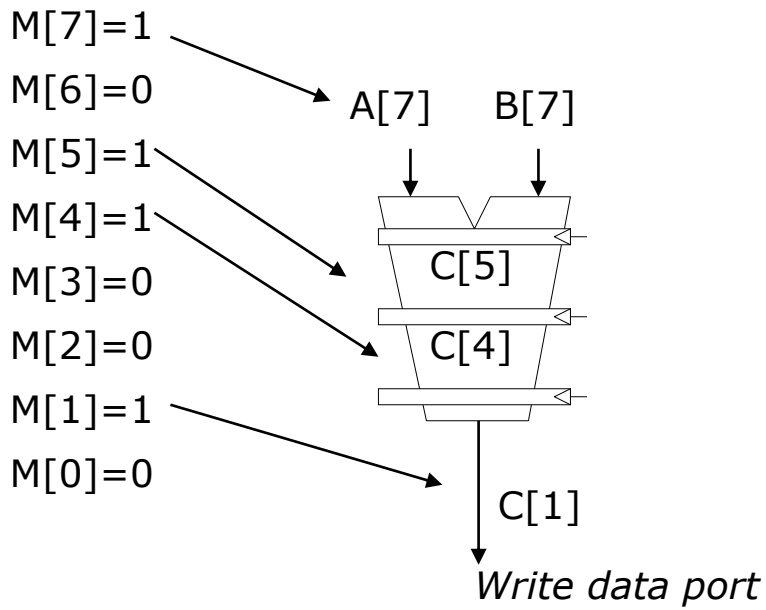
Simple Implementation

- execute all N operations, turn off result writeback according to mask



Density-Time Implementation

- scan mask vector and only execute elements with non-zero masks



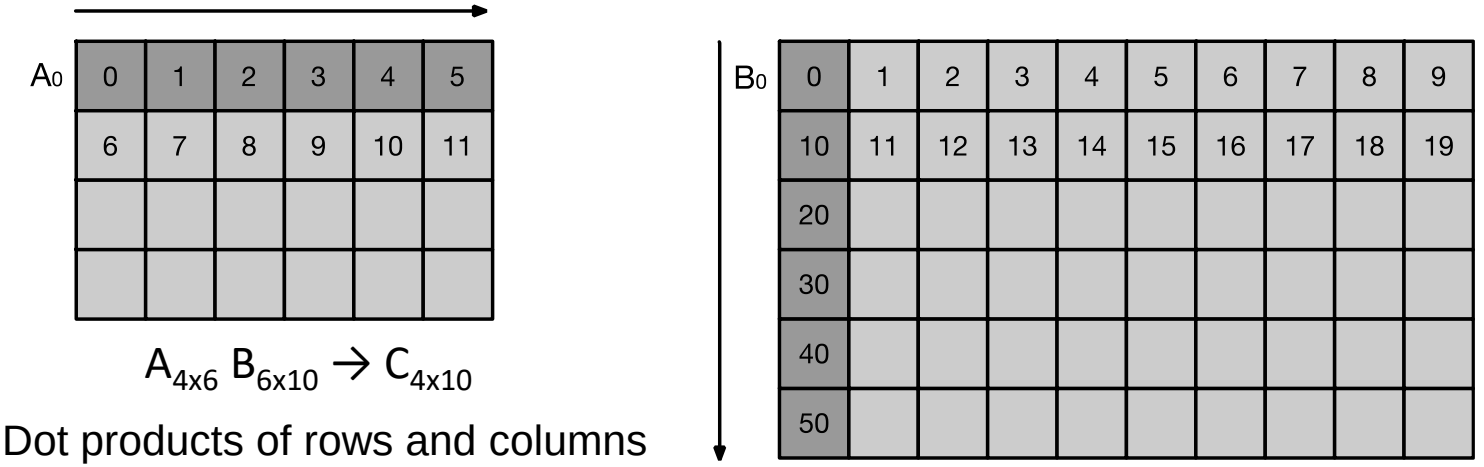
Which one is better?
Tradeoffs?

Some Issues

- Stride and banking
  - As long as they are *relatively prime* to each other and there are enough banks to cover bank access latency, we can sustain 1 element/cycle throughput
- Storage of a matrix
  - Row major: Consecutive elements in a row are laid out consecutively in memory
  - Column major: Consecutive elements in a column are laid out consecutively in memory
  - You need to change the stride when accessing a row versus column

Matrix Multiplication

- A and B, both in row-major order



$A_{4x6}\ B_{6x10} \rightarrow C_{4x10}$

Dot products of rows and columns of A and B

- A: Load A_0 into vector register V_1
  - Each time, increment address by one to access the next column
  - Accesses have a stride of 1
- B: Load B_0 into vector register V_2
  - Each time, increment address by 10
  - Accesses have a stride of 10

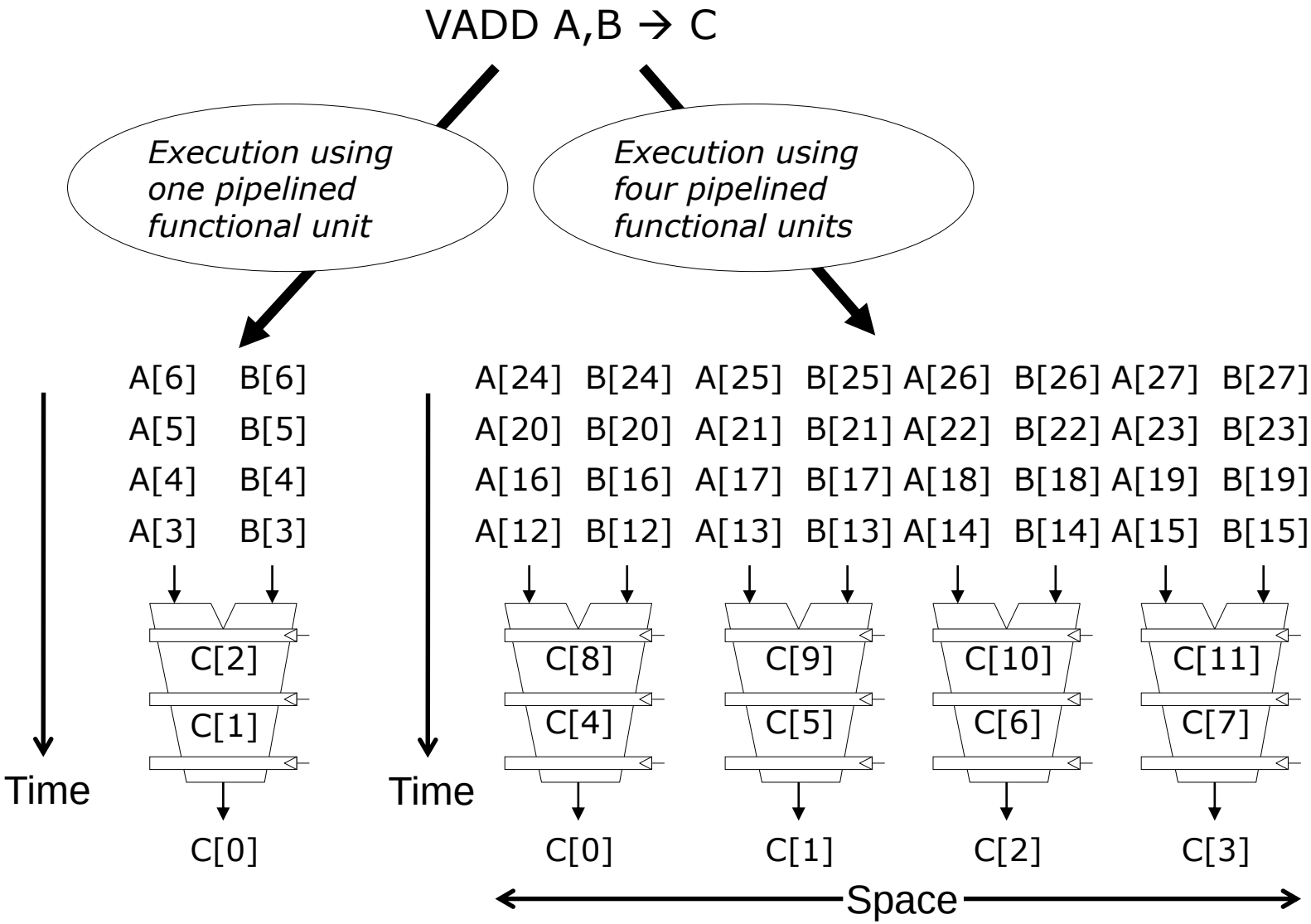
Different strides can lead to bank conflicts

How do we minimize them?

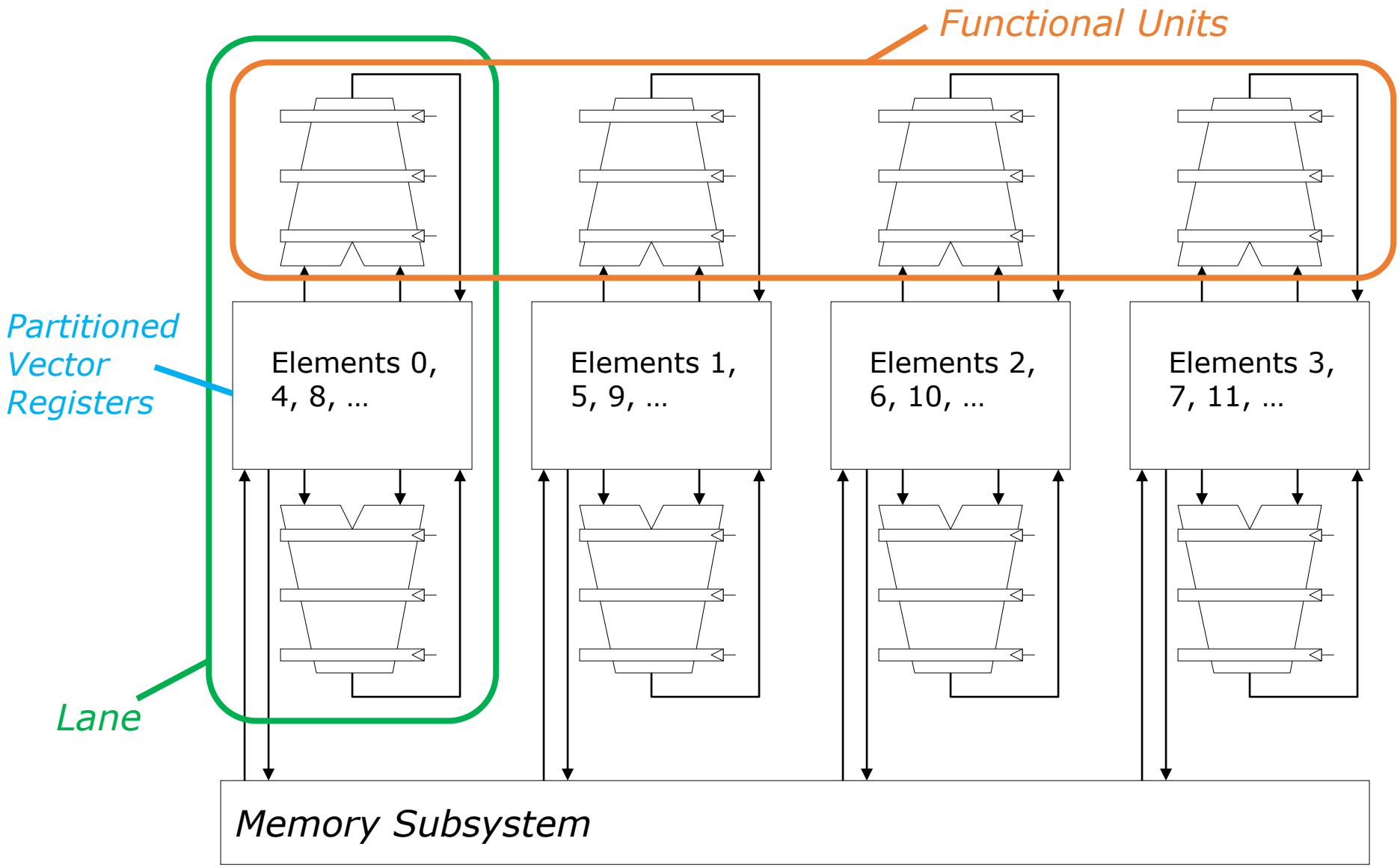
Minimizing Bank Conflicts

- More banks
- Better data layout to match the access pattern
  - Is this always possible?
- Better mapping of address to bank
  - E.g., randomized mapping
  - Rau, "Pseudo-randomly interleaved memory," ISCA 1991.

Vector Instruction Execution



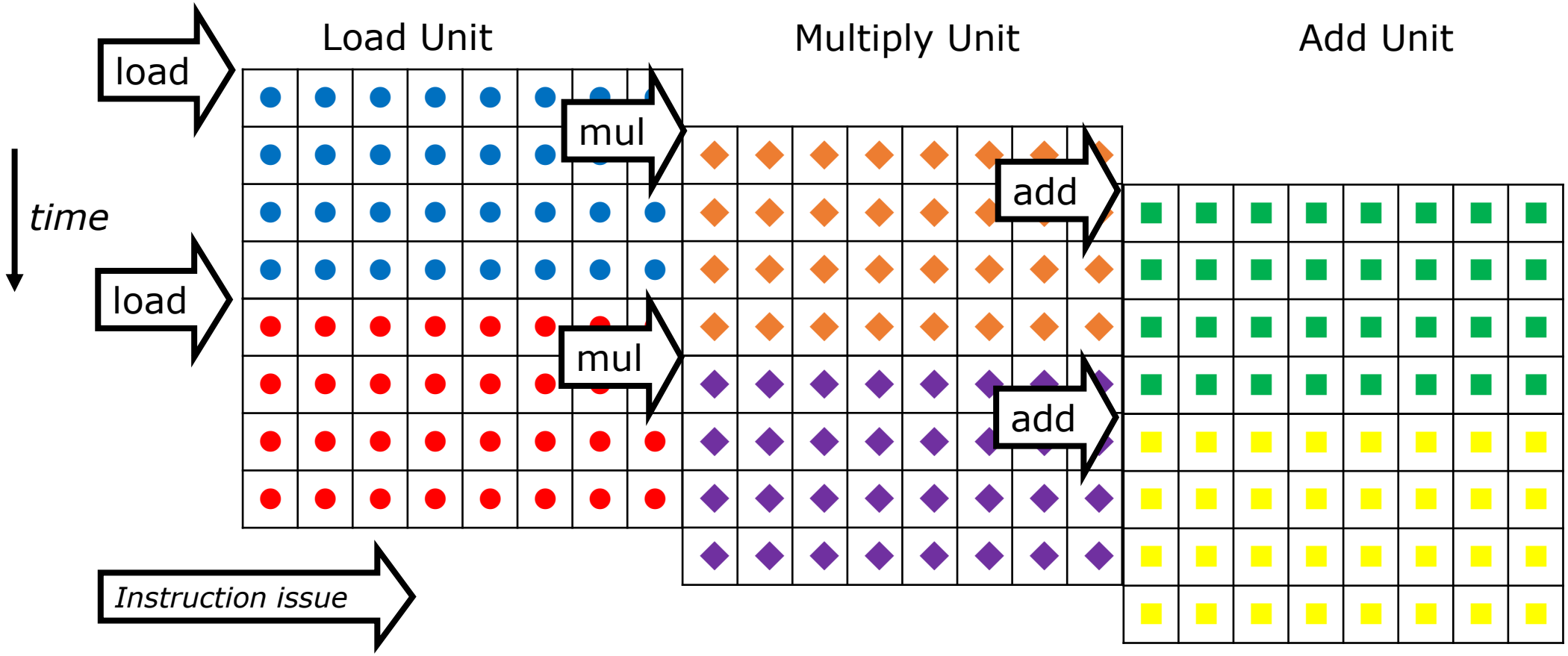
Vector Unit Structure



Vector Instruction Level Parallelism

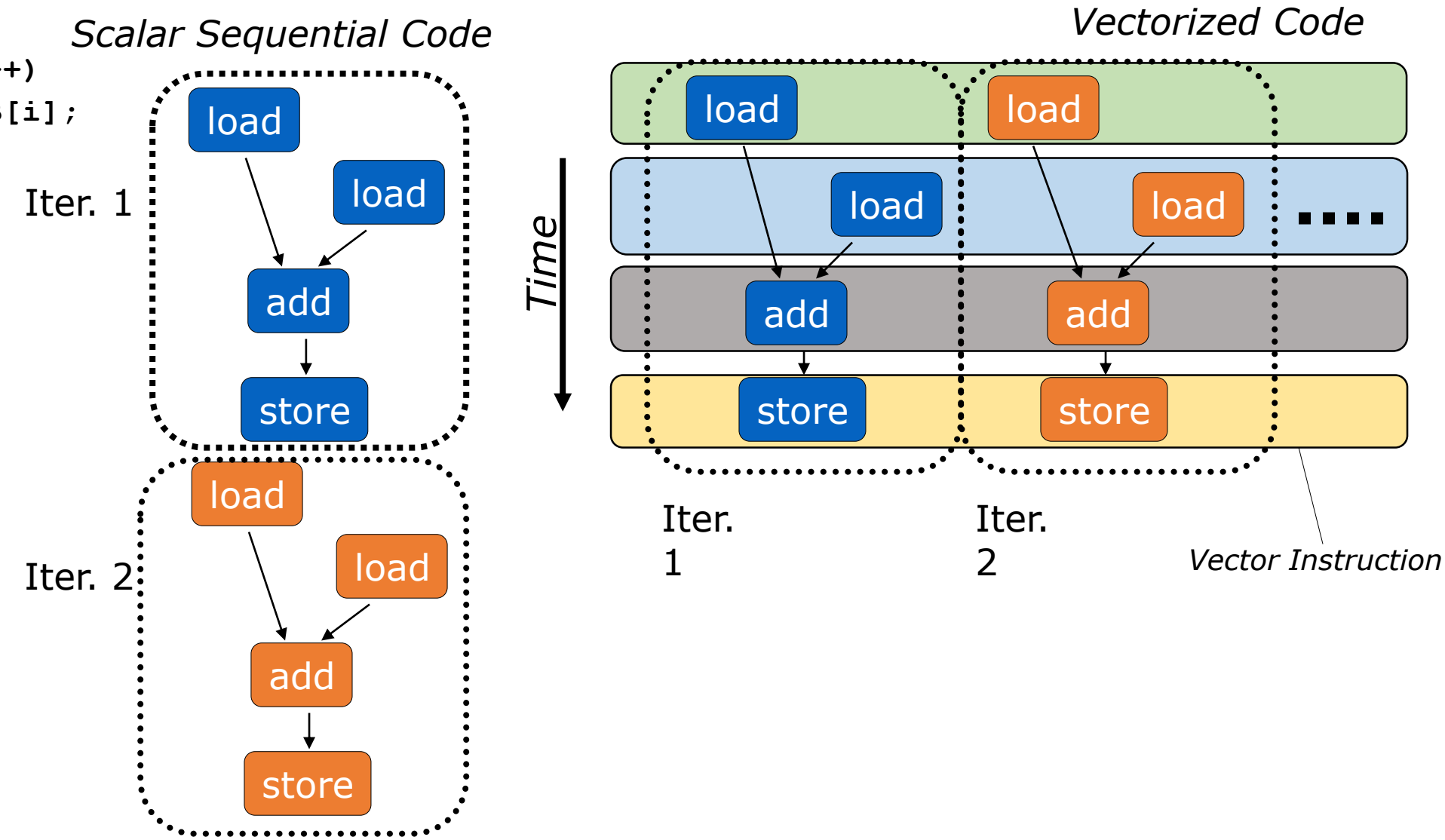
Can overlap execution of multiple vector instructions

- Example machine has 32 elements per vector register and 8 lanes
- Completes 24 operations/cycle while issuing 1 vector instruction/cycle



Automatic Code Vectorization

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



Vectorization is a compile-time reordering of operation sequencing
⇒ requires extensive loop dependence analysis

Vector/SIMD Processing Summary

- Vector/SIMD machines are good at exploiting regular data-level parallelism
  - Same operation performed on many data elements
  - Improve performance, simplify design (no intra-vector dependencies)

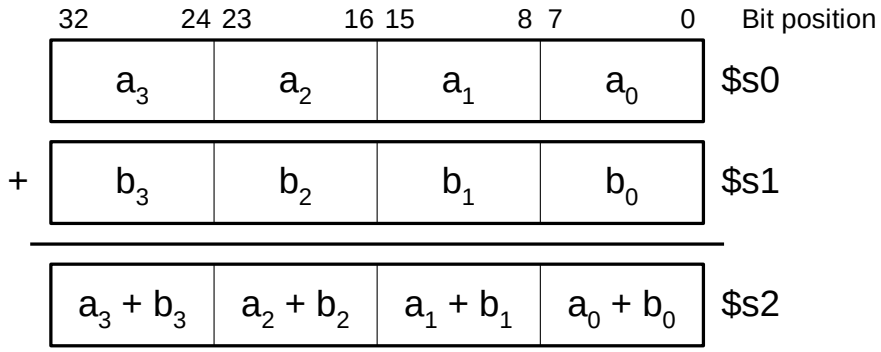
- Performance improvement limited by vectorizability of code
  - Scalar operations limit vector machine performance
  - Remember Amdahl's Law
  - CRAY-1 was the fastest SCALAR machine at its time!

- Many existing ISAs include (vector-like) SIMD operations
  - Intel MMX/SSEn/AVX, PowerPC AltiVec
  - ARM Advanced SIMD/NEON & SVE, RISC-V Vector Extension

SIMD ISA Extensions

- Single Instruction Multiple Data (SIMD) extension instructions
  - Single instruction acts on multiple pieces of data at once
  - Common application: graphics
  - Perform short arithmetic operations (also called packed arithmetic)
- For example: add four 8-bit numbers
- Must modify ALU to eliminate carries between 8-bit values

```
padd8 $s2, $s0, $s1
```



Intel Pentium MMX Operations

- Idea: One instruction operates on multiple data elements simultaneously
  - Designed with multimedia (graphics) operations in mind

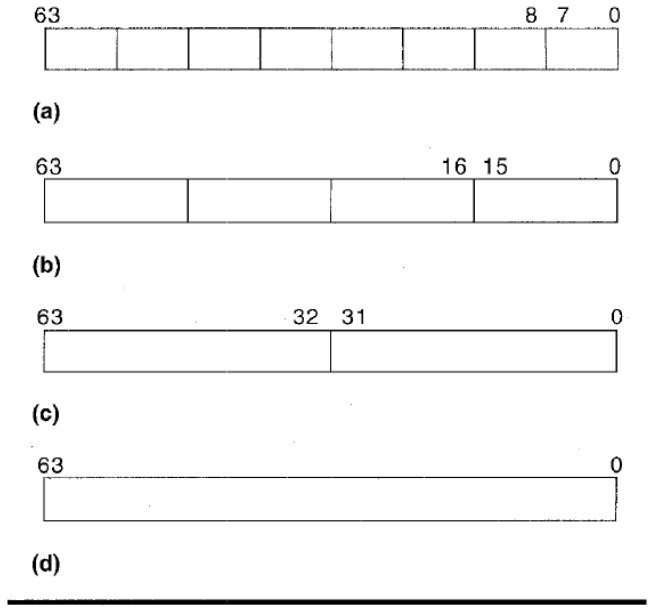


Figure 1. MMX technology data types: packed byte (a), packed word (b), packed doubleword (c), and quadword (d).

No VLEN register
Opcode determines data type:
8 8-bit bytes
4 16-bit words
2 32-bit doublewords
1 64-bit quadword

Stride is always equal to 1.

Peleg and Weiser, "MMX Technology Extension to the Intel Architecture," IEEE Micro, 1996.

Vector Extensions for ARM & RISC-V

- ARM Scalable Vector Extension (SVE)
  - https://developer.arm.com/solutions/hpc/resources/hpc-white-papers/arm-scalable-vector-extensions-and-application-to-machine-learning
  - https://gitlab.com/arm-hpc/training/bsc_training_materials/-/blob/master/Slides/7%20-%20Vectorization%20with%20SVE.pptx

- RISC-V Vector Extensions
  - https://github.com/riscv/riscv-v-spec/blob/master/v-spec.adoc
  - https://riscv.org//wp-content/uploads/2019/06/17.40-Vector_RISCV-20190611-Vectors.pdf