

CS 152 Computer Architecture and Engineering

CS252 Graduate Computer Architecture

Lecture 15 – Vectors

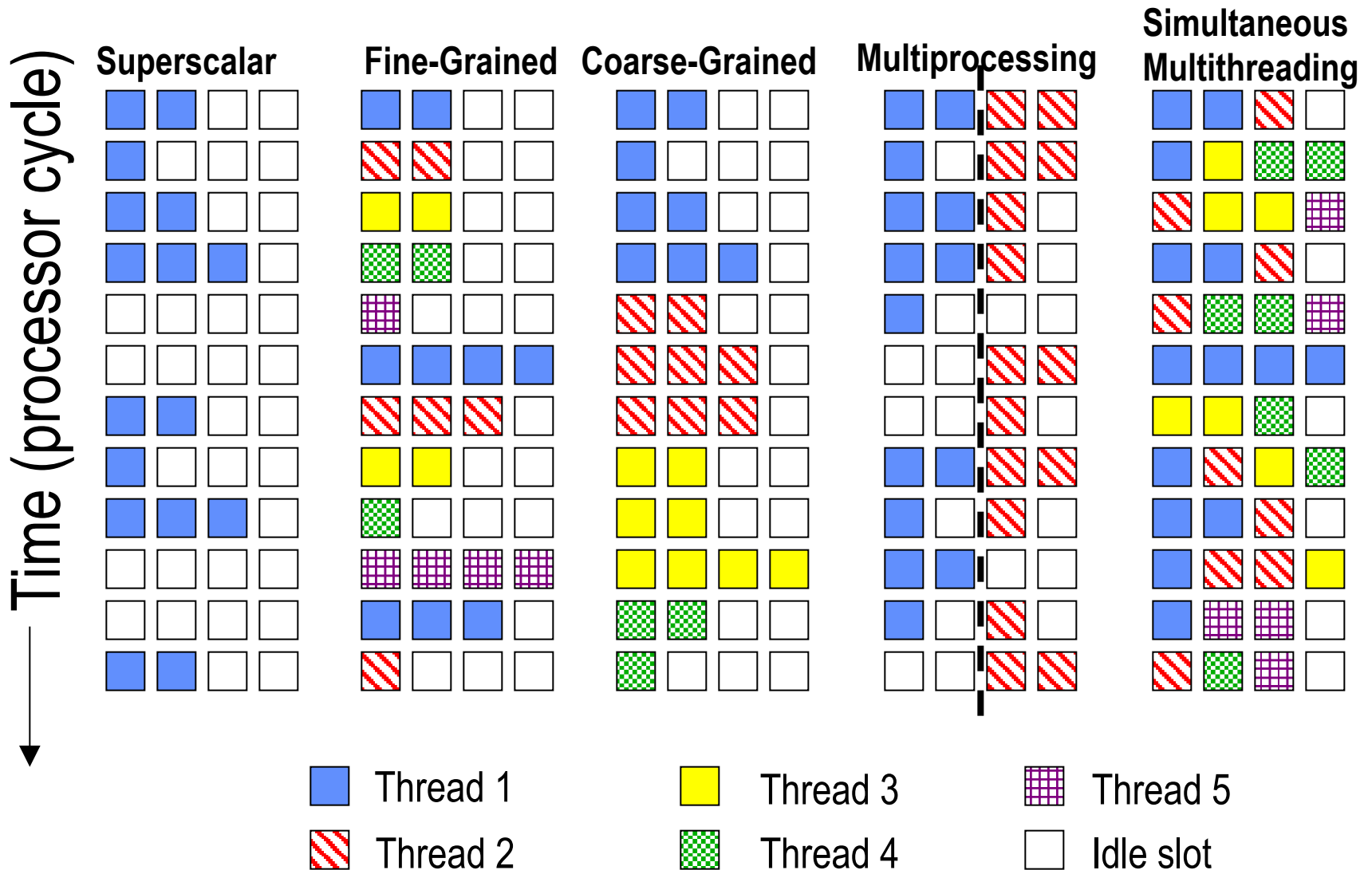
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`https://cwfletcher.github.io/`

`http://inst.eecs.berkeley.edu/~cs152`

Last Time Lecture 14: Multithreading



Supercomputer Applications

- Typical application areas
 - Military research (nuclear weapons, cryptography)
 - Scientific research
 - Weather forecasting
 - Oil exploration
 - Industrial design (car crash simulation)
 - Bioinformatics
 - Cryptography
- All involve huge computations on large data set
- Supercomputers: CDC6600, CDC7600, Cray-1, ...
- In 70s-80s, Supercomputer \equiv Vector Machine

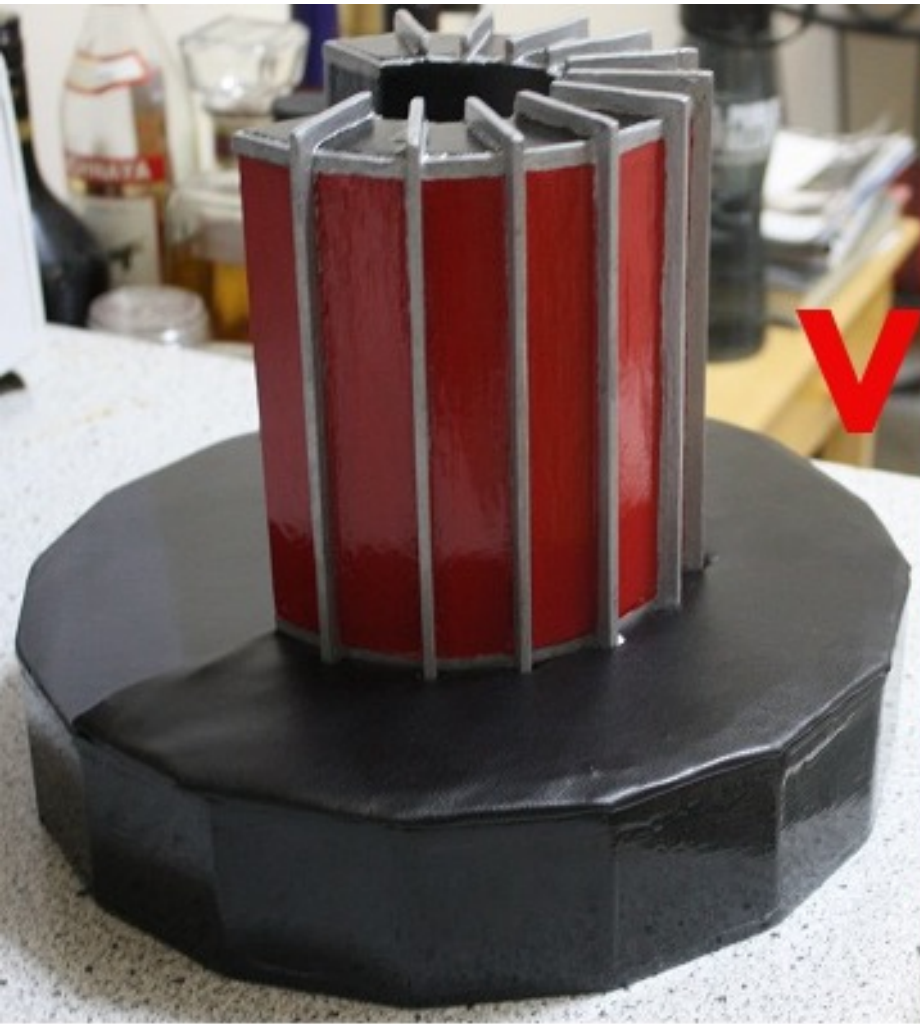
Vector Supercomputers



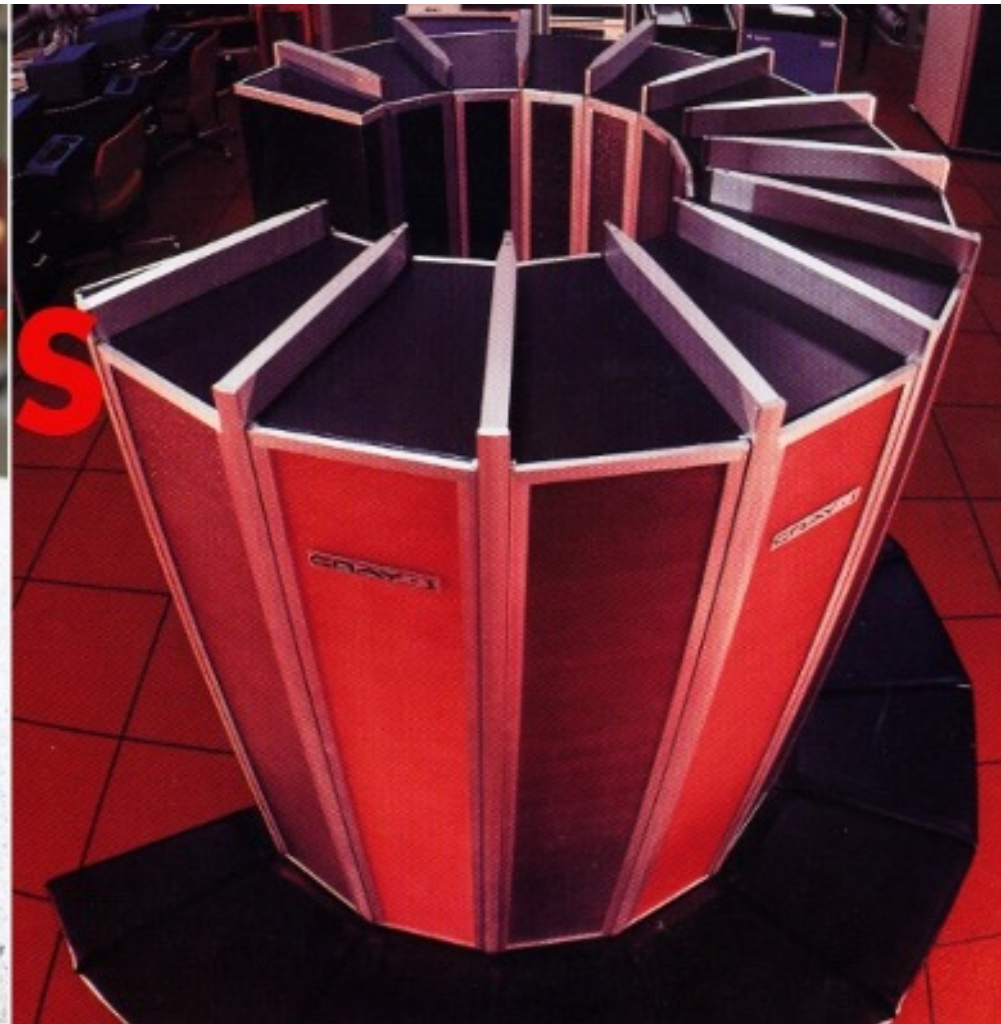
- Epitomized by Cray-1, 1976:
- Scalar Unit
 - Load/Store Architecture
- Vector Extension
 - Vector Registers
 - Vector Instructions
- Implementation
 - Hardwired Control
 - Highly Pipelined Functional Units
 - Interleaved Memory System
 - No Data Caches
 - No Virtual Memory

[©Cray Research, 1976]

Cray-1 today



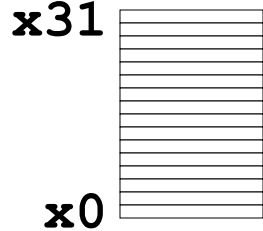
VS



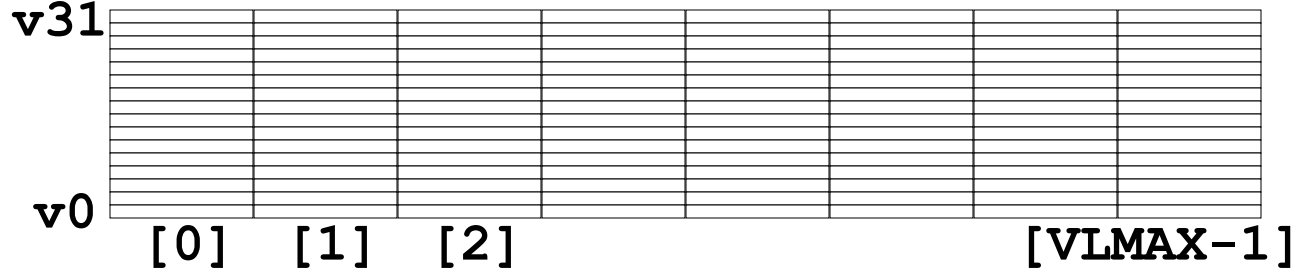
<https://www.chrisfenton.com/homebrew-cray-1a/>

Vector Programming Model

Scalar Registers



Vector Registers

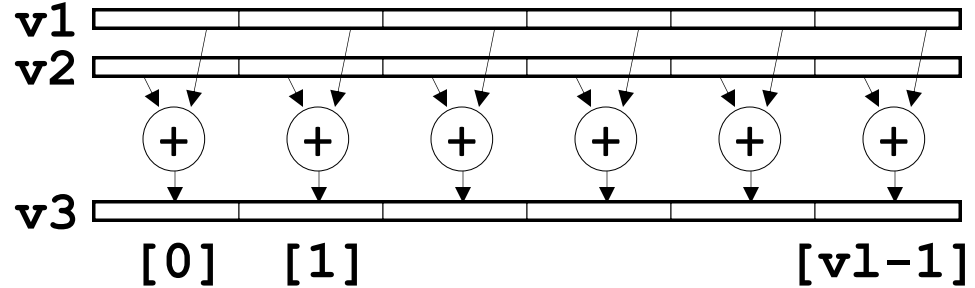


Vector Length Register



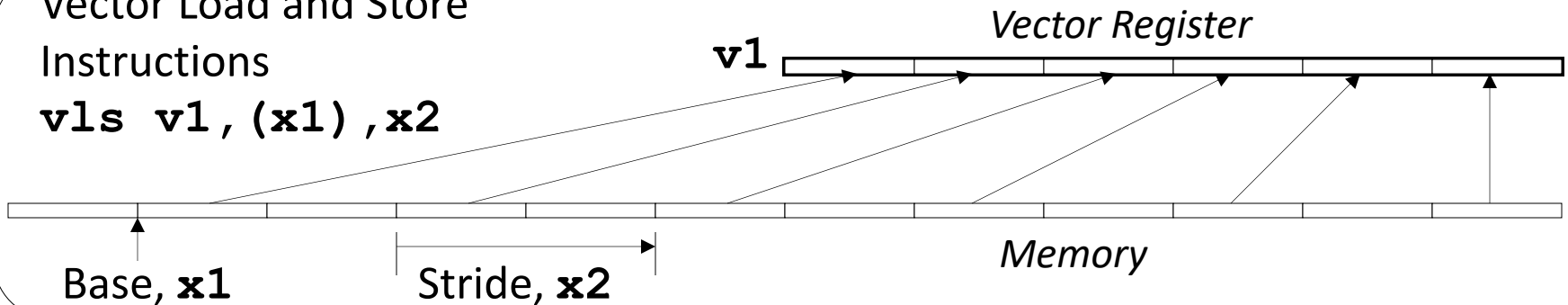
Vector Arithmetic Instructions

`vadd v3, v1, v2`



Vector Load and Store Instructions

`vls v1, (x1), x2`



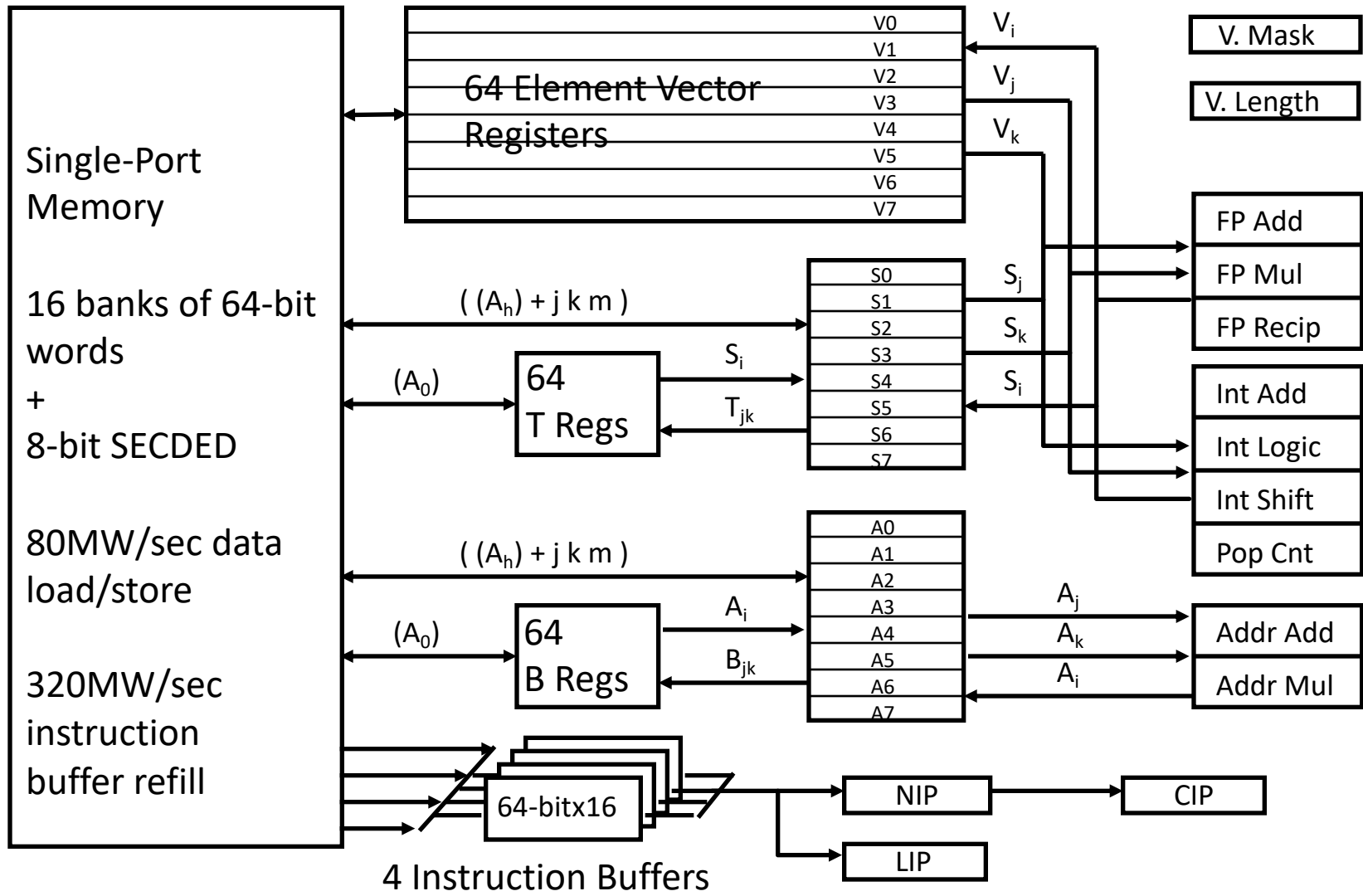
Vector Code Example

```
# C code
for (i=0; i<64; i++)
    C[i] = A[i] + B[i];
```

```
# Scalar Code
    li x4, 64
loop:
    fld f1, 0(x1)
    fld f2, 0(x2)
    fadd.d f3, f1, f2
    fsd f3, 0(x3)
    addi x1, x1, 8
    addi x2, x2, 8
    addi x3, x3, 8
    subi x4, x4, 1
    bnez x4, loop
```

```
# Vector Code
    li x4, 64
    vsetv1 x4
    vld v1, (x1)
    vld v2, (x2)
    vadd v3, v1, v2
    vst v3, (x3)
```

Cray-1 (1976)



memory bank cycle 50 ns processor cycle 12.5 ns (80MHz)

Vector Instruction Set Advantages

- **Compact**

- one short instruction encodes N operations

- **Expressive, tells hardware that these N operations:**

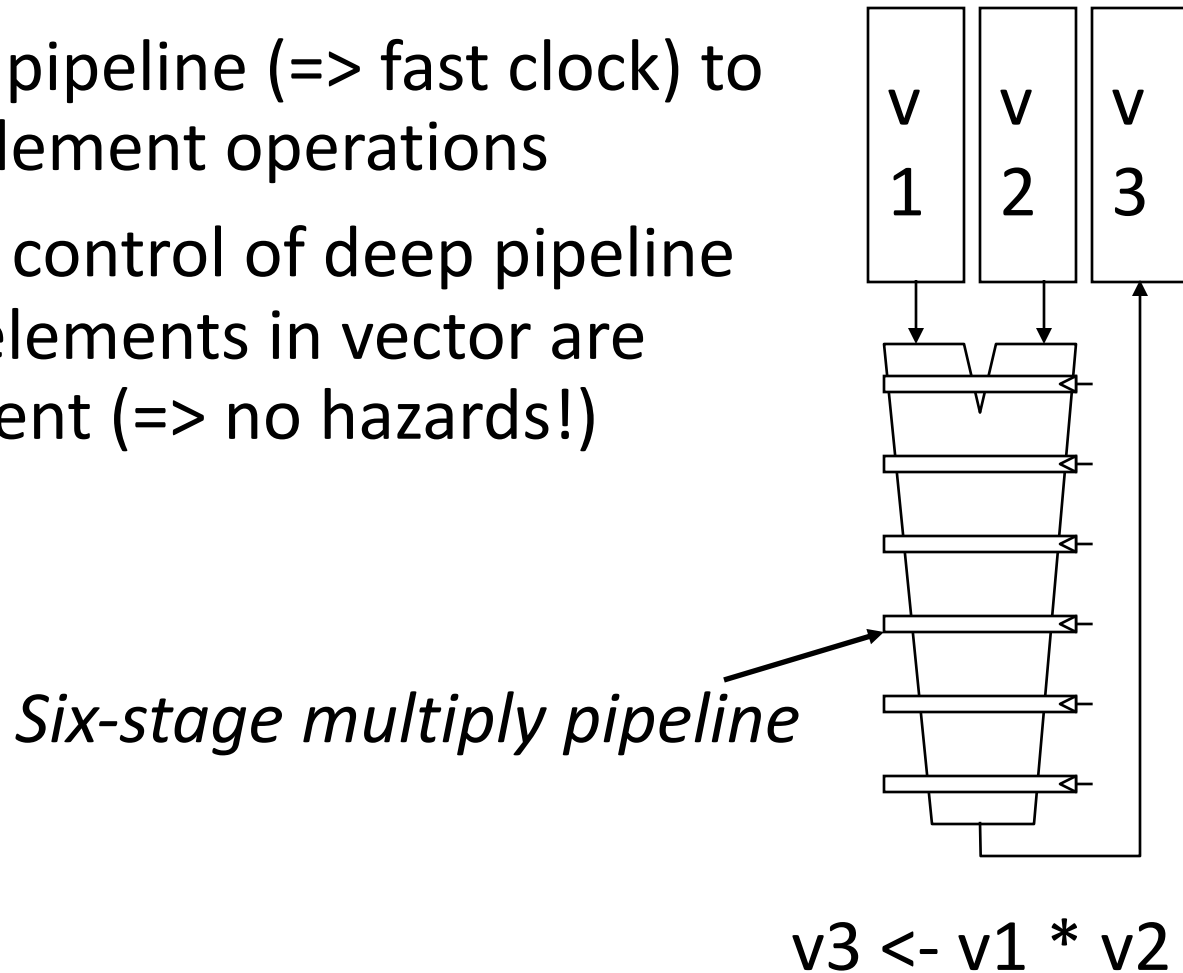
- are independent
- use the same functional unit
- access disjoint registers
- access registers in same pattern as previous instructions
- access a contiguous block of memory (unit-stride load/store)
- access memory in a known pattern (strided load/store)

- **Scalable**

- can run same code on more parallel pipelines (lanes)

Vector Arithmetic Execution

- Use deep pipeline (\Rightarrow fast clock) to execute element operations
- Simplifies control of deep pipeline because elements in vector are independent (\Rightarrow no hazards!)



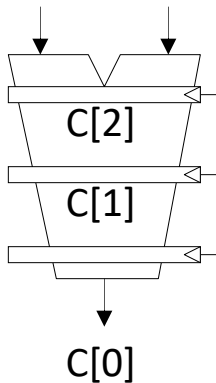
Vector Instruction Execution

vadd vc, va, vb

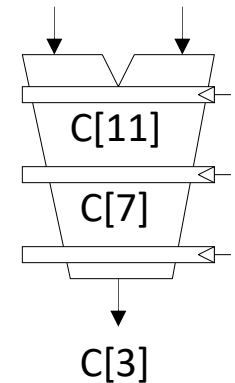
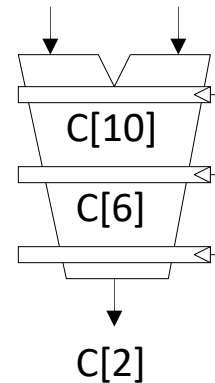
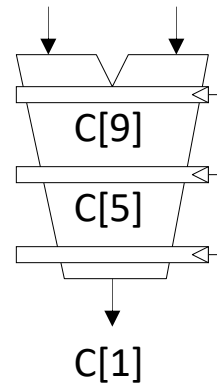
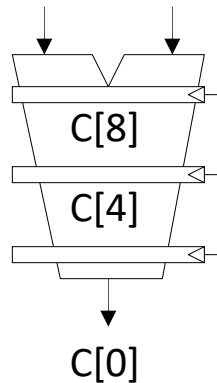
*Execution using
one pipelined
functional unit*

*Execution using
four pipelined
functional units*

A[6] B[6]
A[5] B[5]
A[4] B[4]
A[3] B[3]

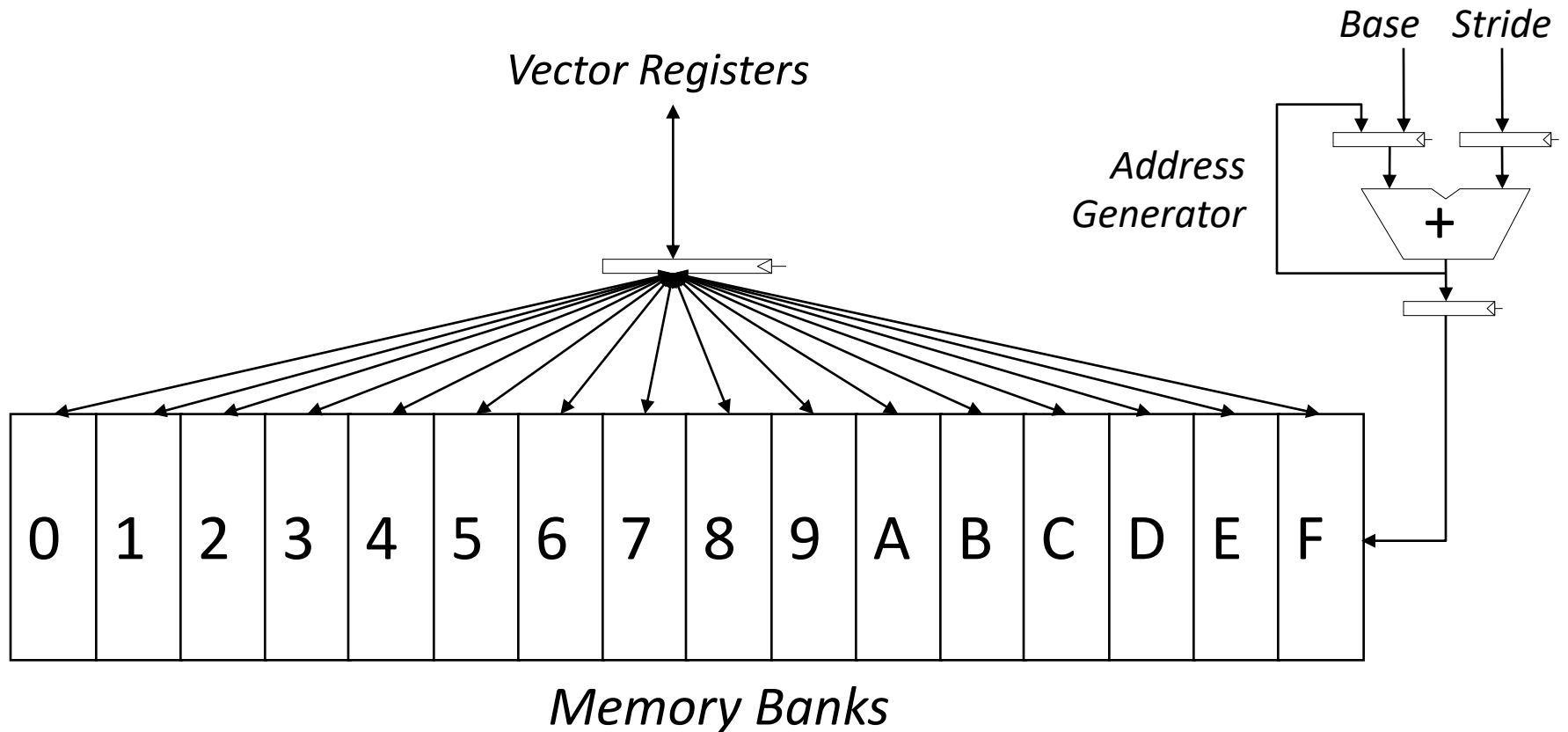


A[24] B[24] A[25] B[25] A[26] B[26] A[27] B[27]
A[20] B[20] A[21] B[21] A[22] B[22] A[23] B[23]
A[16] B[16] A[17] B[17] A[18] B[18] A[19] B[19]
A[12] B[12] A[13] B[13] A[14] B[14] A[15] B[15]

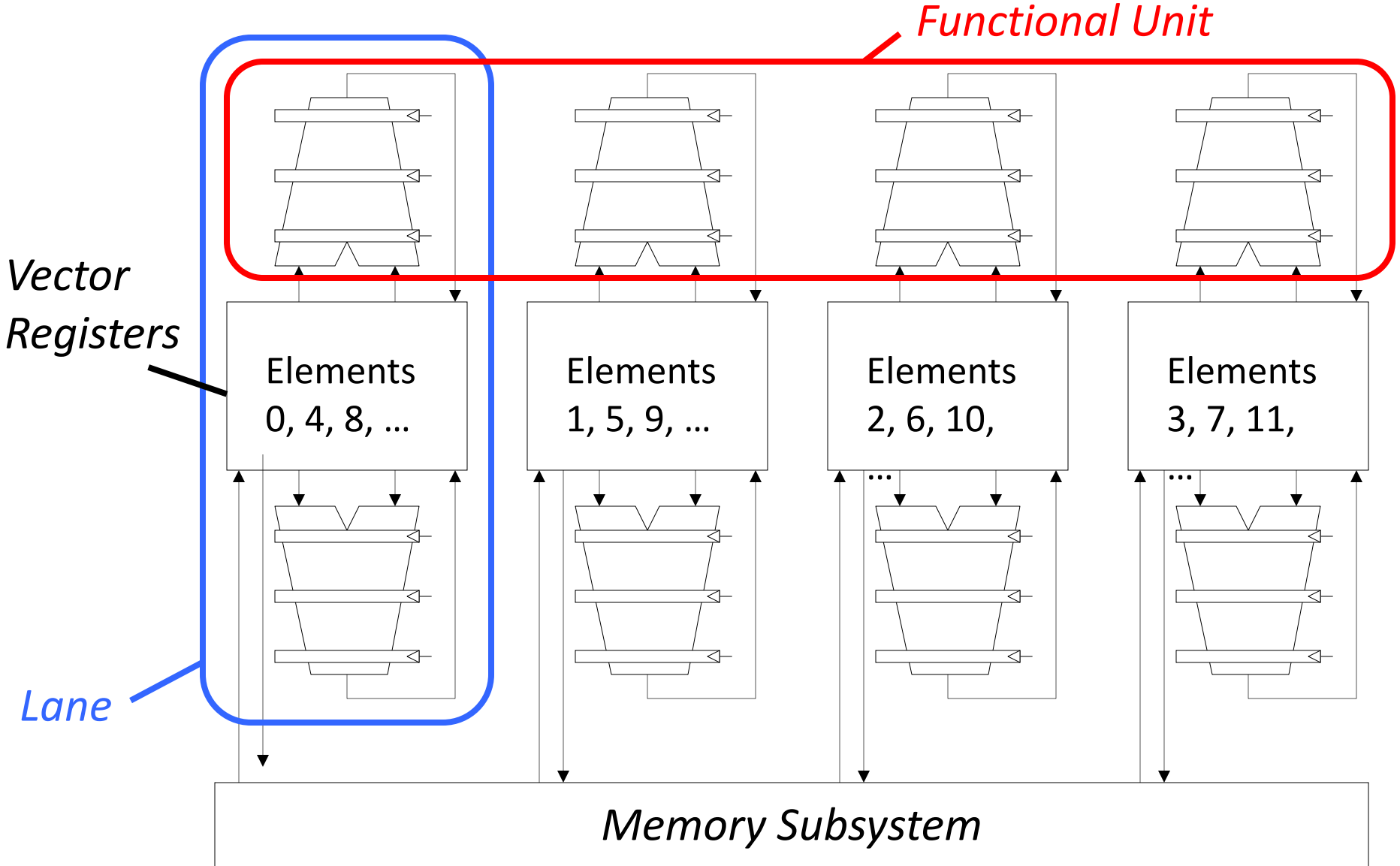


Interleaved Vector Memory System

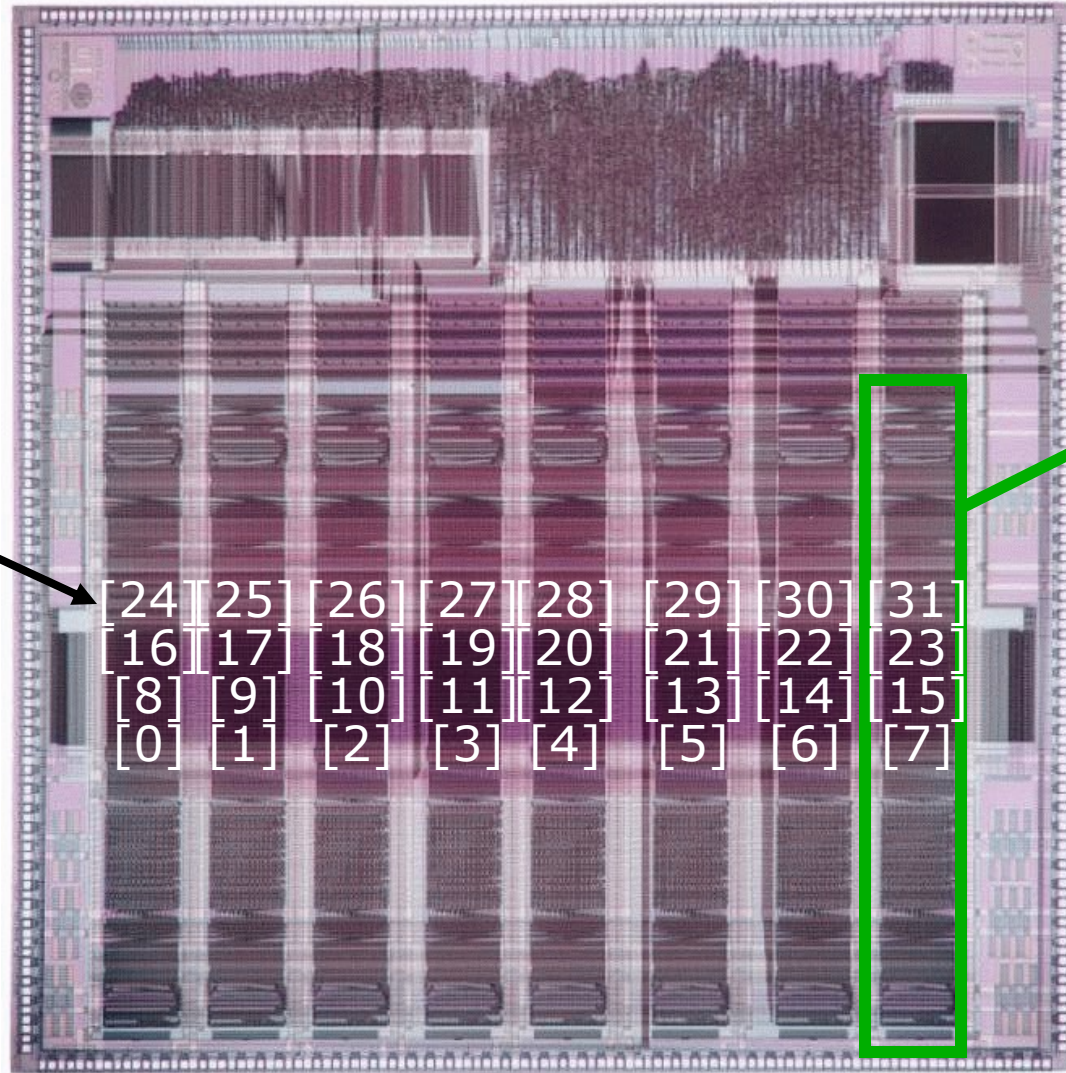
- Bank busy time: Time before bank ready to accept next request
- Cray-1, 16 banks, 4 cycle bank busy time, 12 cycle latency



Vector Unit Structure



T0 Vector Microprocessor (UCB/ICSI, 1995)



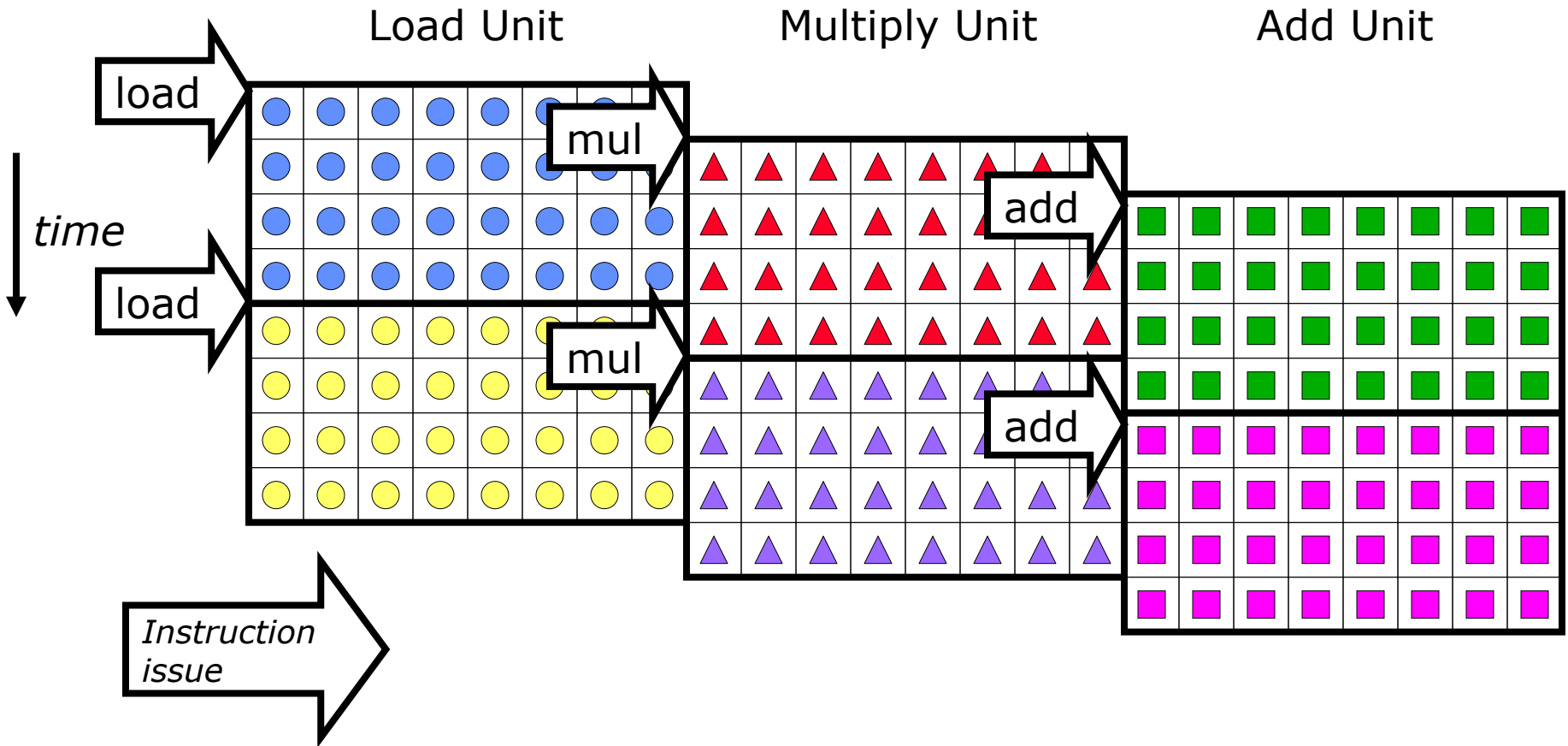
Vector register elements striped over lanes

Lane

[24]	[25]	[26]	[27]	[28]	[29]	[30]	[31]
[16]	[17]	[18]	[19]	[20]	[21]	[22]	[23]
[8]	[9]	[10]	[11]	[12]	[13]	[14]	[15]
[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]

Vector Instruction Parallelism

- Can overlap execution of multiple vector instructions
 - example machine has 32 elements per vector register and 8 lanes

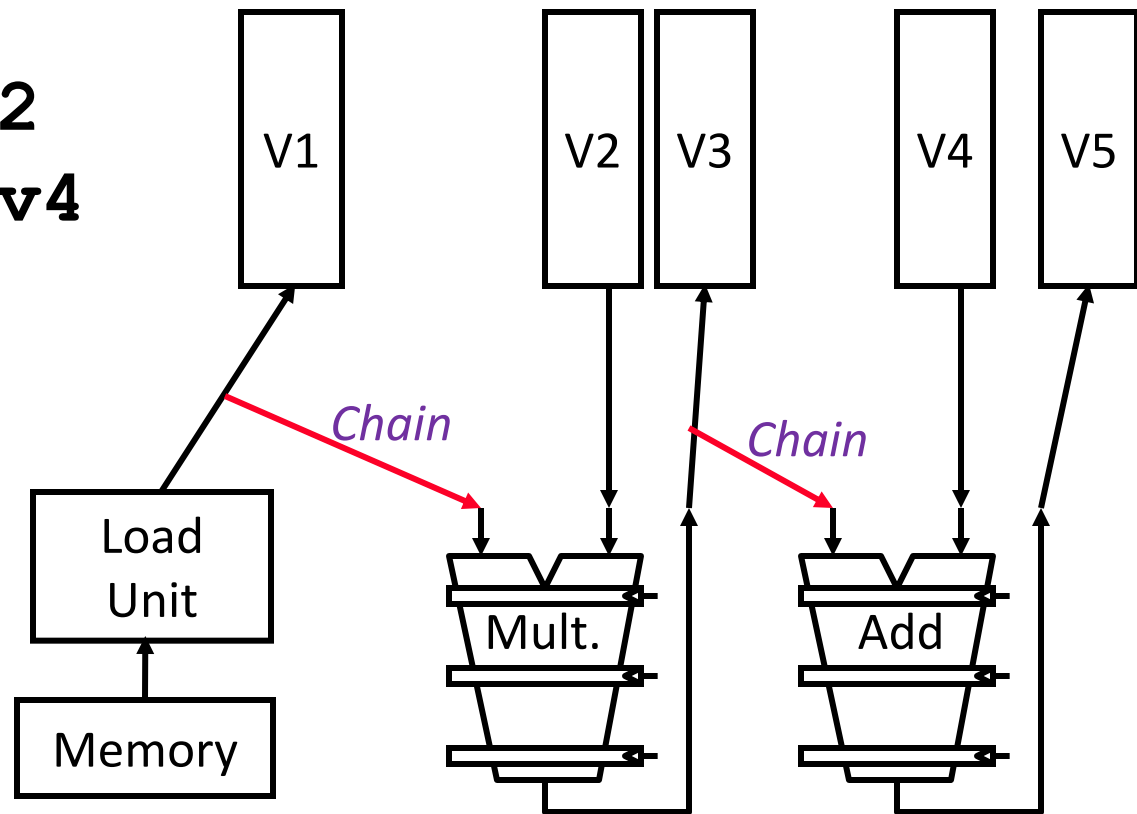


Complete 24 operations/cycle while issuing 1 short instruction/cycle

Vector Chaining

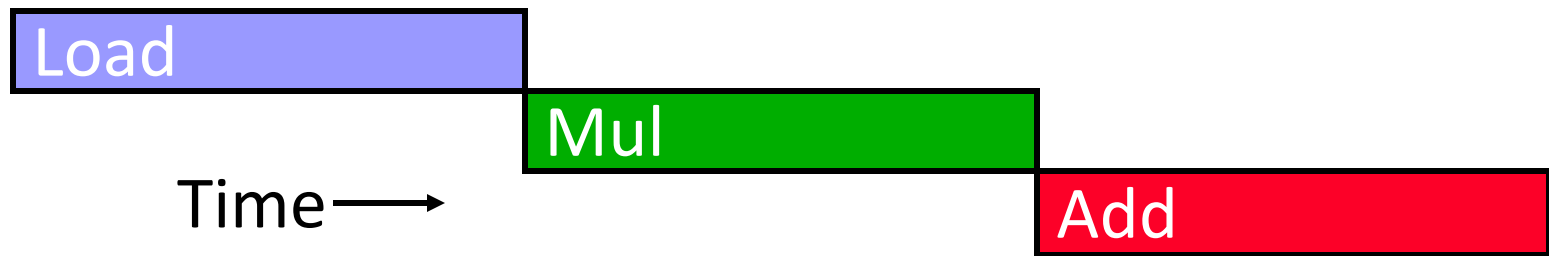
- Vector version of register bypassing
 - introduced with Cray-1

```
vld  v1  
vmul v3, v1, v2  
vadd v5, v3, v4
```



Vector Chaining Advantage

- Without chaining, must wait for last element of result to be written before starting dependent instruction



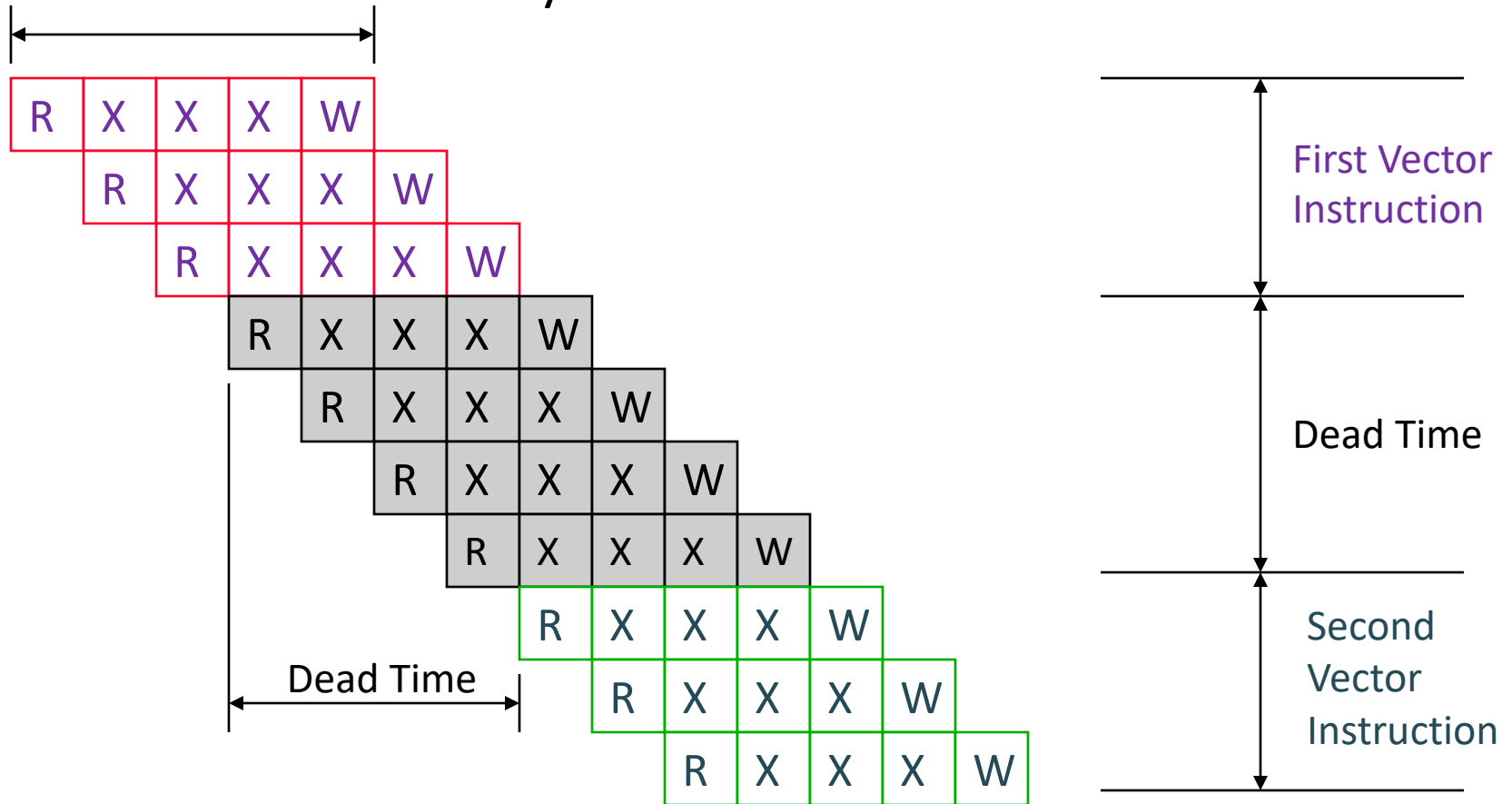
- With chaining, can start dependent instruction as soon as first result appears



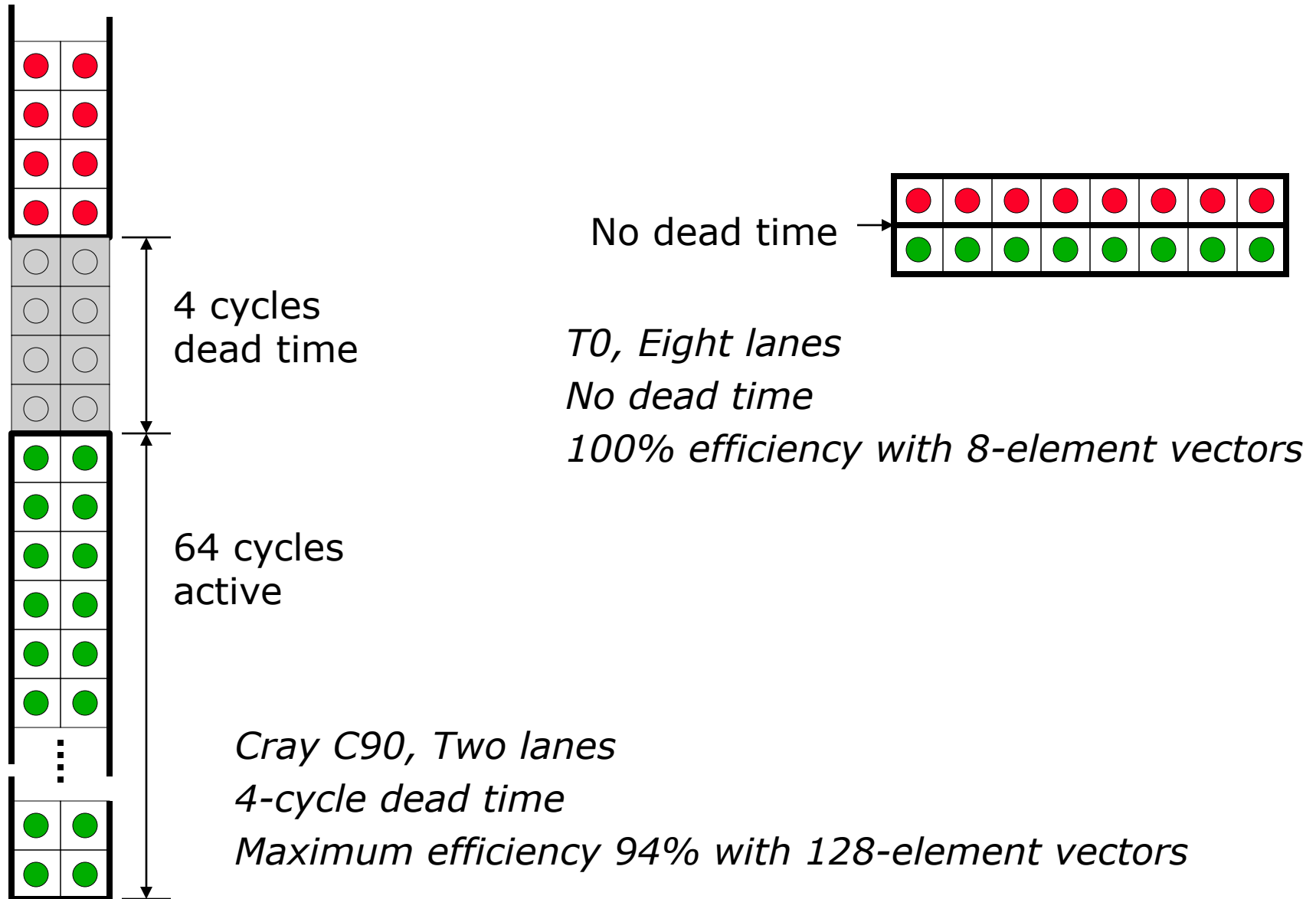
Vector Startup

- Two components of vector startup penalty
 - functional unit latency (time through pipeline)
 - dead time or recovery time (time before another vector instruction can start down pipeline)

Functional Unit Latency



Dead Time and Short Vectors



Vector Memory-Memory versus Vector Register Machines

- Vector memory-memory instructions hold all vector operands in main memory
- The first vector machines, CDC Star-100 ('73) and TI ASC ('71), were memory-memory machines
- Cray-1 ('76) was first vector register machine

Example Source Code

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

Vector Memory-Memory Code

```
vadd (C) , (A) , (B)  
vsub (D) , (A) , (B)
```

Vector Register Code

```
vld V1, (A)  
vld V2, (B)  
vadd V3, V1, V2  
vst V3, (C)  
vsub V4, V1, V2  
vst V4, (D)
```

Vector Memory-Memory vs. Vector Register Machines

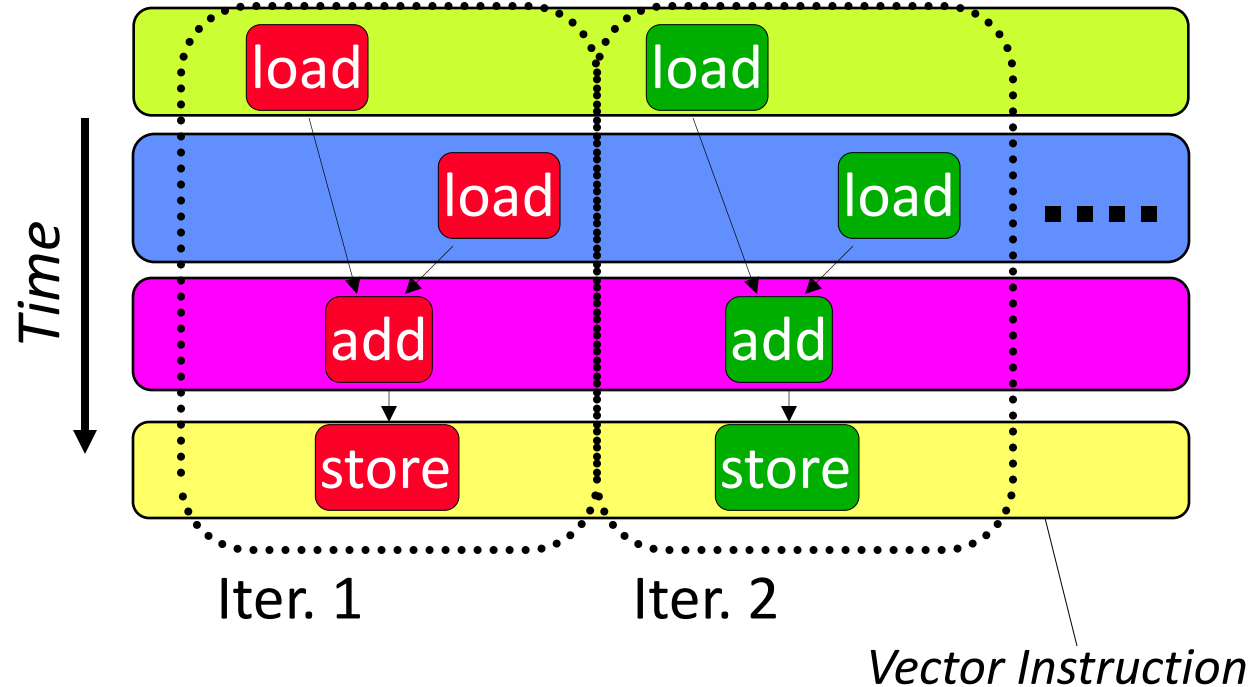
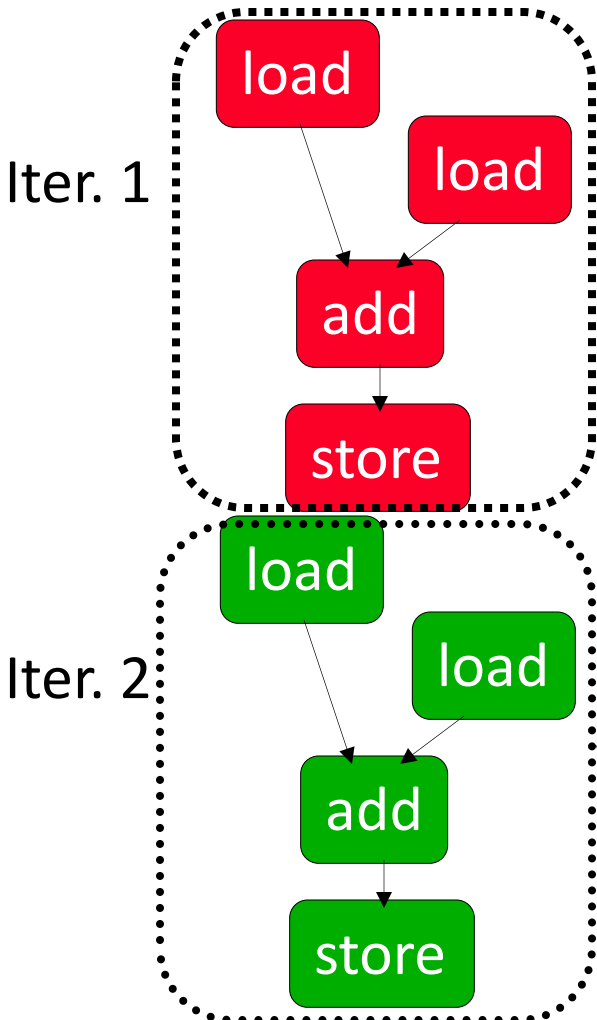
- Vector memory-memory architectures (VMMA) require greater main memory bandwidth, why?
 - All operands must be read in and out of memory
- VMMA make it difficult to overlap execution of multiple vector operations, why?
 - Must check dependencies on memory addresses
- VMMA incur greater startup latency
 - Scalar code was faster on CDC Star-100 for vectors < 100 elements
 - For Cray-1, vector/scalar breakeven point was around 2-4 elements
- Apart from CDC follow-ons (Cyber-205, ETA-10) all major vector machines since Cray-1 have had vector register architectures
- (we ignore vector memory-memory from now on)

Automatic Code Vectorization

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

Scalar Sequential Code

Vectorized Code



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

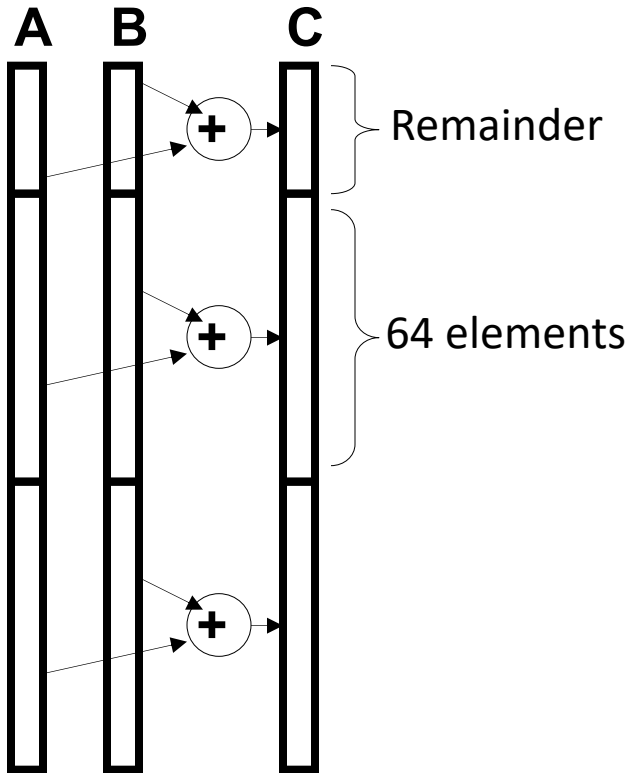
Vector Stripmining

Problem: Vector registers have finite length

Solution: Break loops into pieces that fit in registers, “Stripmining”

```
andi x1, xN, 63 # N mod 64
vsetvl x1      # Do remainder
loop:
vld v1, (xA)
slli x2, x1, 3 # Multiply by 8
add xA, xA, x2 # Bump pointer
vld v2, (xB)
add xB, xB, x2
vadd v3, v1, v2
vst v3, (xC)
add xC, xC, x2
sub xN, xN, x1 # Subtract elements
li x1, 64
vsetvl x1      # Reset full length
bgtz xN, loop # Any more to do?
```

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



Vector Conditional Execution

Problem: Want to vectorize loops with conditional code:

```
for (i=0; i<N; i++)  
    if (A[i]>0) then  
        A[i] = B[i];
```

Solution: Add vector *mask* (or *flag*) registers

- vector version of predicate registers, 1 bit per element

...and *maskable* vector instructions

- vector operation becomes bubble (“NOP”) at elements where mask bit is clear

Code example:

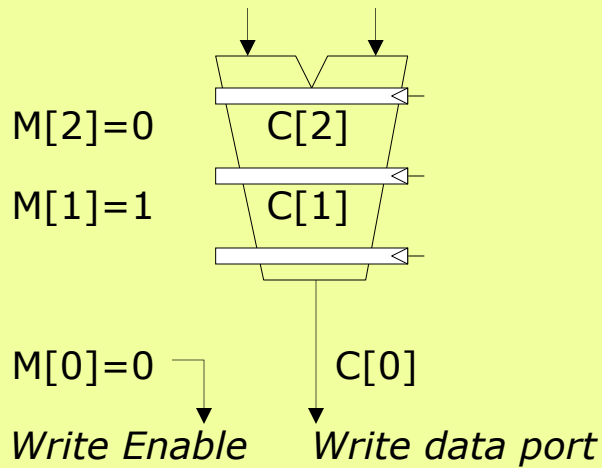
```
cvm                # Turn on all elements  
vld vA, (xA)       # Load entire A vector  
vgt vA, f0         # Set bits in mask register where A>0  
vld vA, (xB)       # Load B vector into A under mask  
vst vA, (xA)       # Store A back to memory under mask
```


Masked Vector Instructions

Simple Implementation

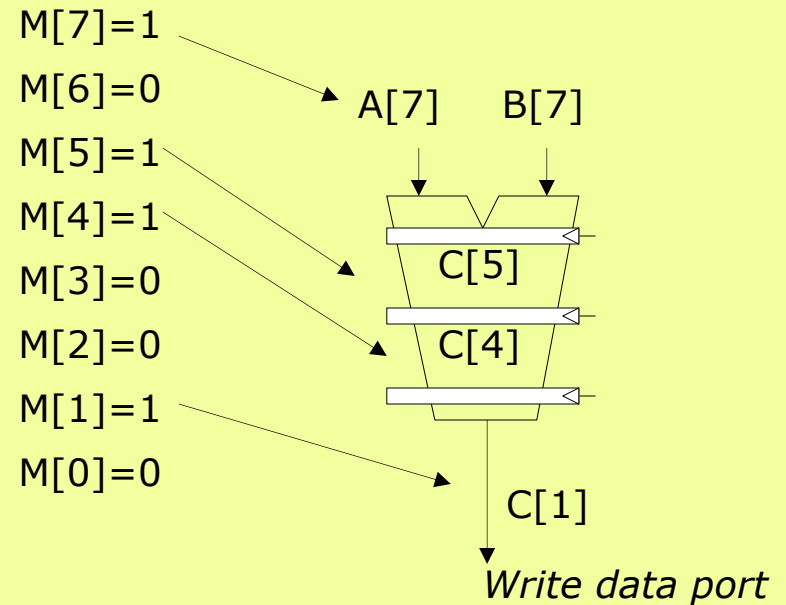
- execute all N operations, turn off result writeback according to mask

M[7]=1 A[7] B[7]
M[6]=0 A[6] B[6]
M[5]=1 A[5] B[5]
M[4]=1 A[4] B[4]
M[3]=0 A[3] B[3]



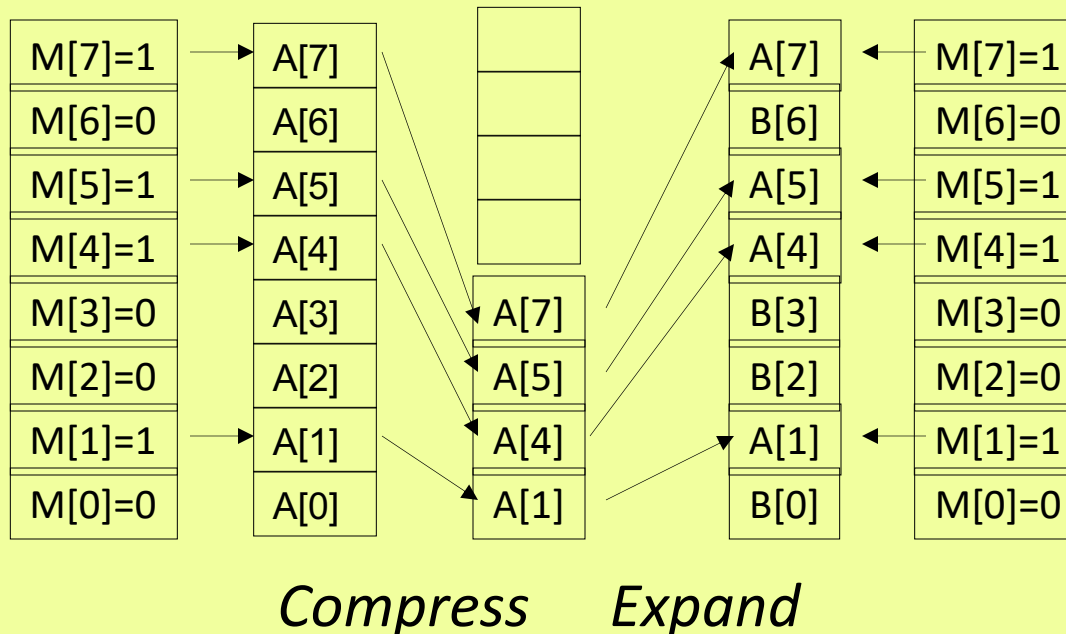
Density-Time Implementation

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



Compress/Expand Operations

- Compress packs non-masked elements from one vector register contiguously at start of destination vector register
 - population count of mask vector gives packed vector length
- Expand performs inverse operation



Used for density-time conditionals and also for general selection operations

Vector Reductions

Problem: Loop-carried dependence on reduction variables

```
sum = 0;
for (i=0; i<N; i++)
    sum += A[i]; # Loop-carried dependence on sum
```

Solution: Re-associate operations if possible, use binary tree to perform reduction

```
# Rearrange as:
sum[0:VL-1] = 0 # Vector of VL partial sums
for(i=0; i<N; i+=VL) # Stripmine VL-sized chunks
    sum[0:VL-1] += A[i:i+VL-1]; # Vector sum
# Now have VL partial sums in one vector register
do {
    VL = VL/2; # Halve vector length
    sum[0:VL-1] += sum[VL:2*VL-1] # Halve no. partials
} while (VL>1)
```

Vector Scatter/Gather

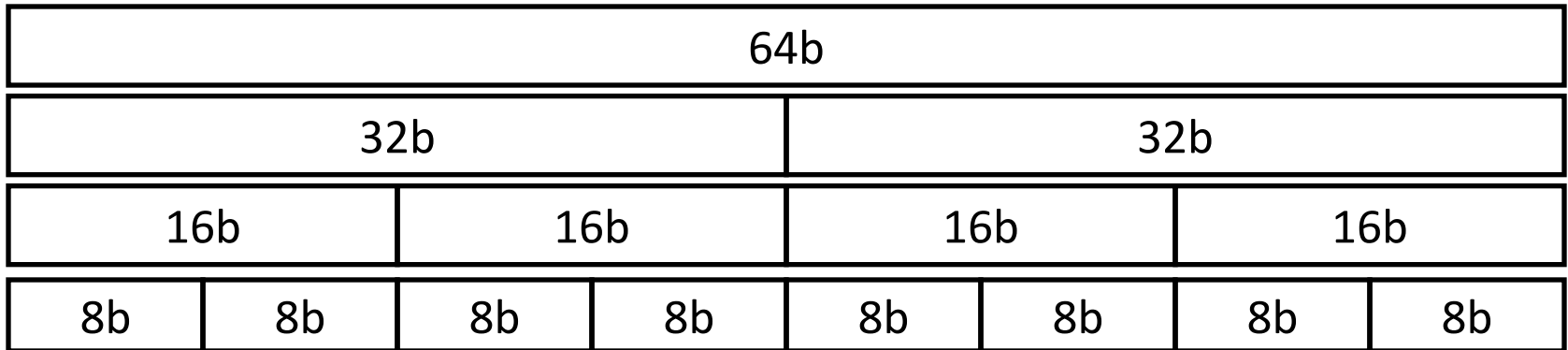
Want to vectorize loops with indirect accesses:

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

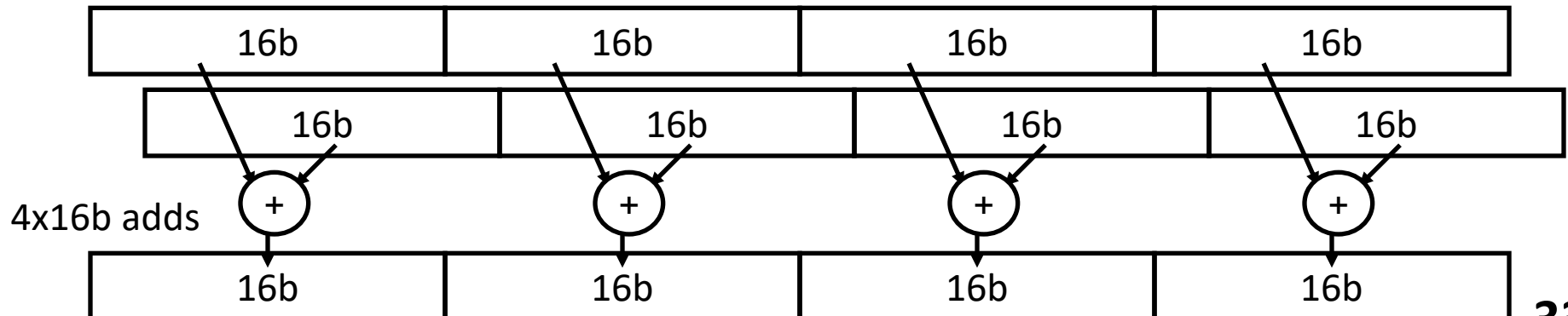
Indexed load instruction (*Gather*)

```
vld vD, (xD)           # Load indices in D vector  
vlx vC, (xC), vD       # Load indexed from xC base  
vld vB, (xB)           # Load B vector  
vadd vA, vB, vC        # Do add  
vst vA, (xA)           # Store result
```

Packed SIMD Extensions



- Very short vectors added to existing ISAs for microprocessors
- Use existing 64-bit registers split into 2x32b or 4x16b or 8x8b
 - Lincoln Labs TX-2 from 1957 had 36b datapath split into 2x18b or 4x9b
 - Newer designs have wider registers
 - 128b for PowerPC AltiVec, Intel SSE2/3/4
 - 256b/512b for Intel AVX
- Single instruction operates on all elements within register



Packed SIMD versus Vectors

- Limited instruction set:
 - no vector length control
 - no strided load/store or scatter/gather
 - unit-stride loads must be aligned to 64/128-bit boundary
- Limited vector register length:
 - requires superscalar dispatch to keep multiply/add/load units busy
 - loop unrolling to hide latencies increases register pressure
- Trend towards fuller vector support in microprocessors
 - Better support for misaligned memory accesses
 - Support of double-precision (64-bit floating-point)
 - New Intel AVX spec (announced April 2008), 256b vector registers (expandable up to 1024b), gather added, scatter to follow
 - ARM Scalable Vector Extensions (SVE)

Acknowledgements

- This course is partly inspired by previous MIT 6.823 and Berkeley CS252 computer architecture courses created by my collaborators and colleagues:
 - Arvind (MIT)
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