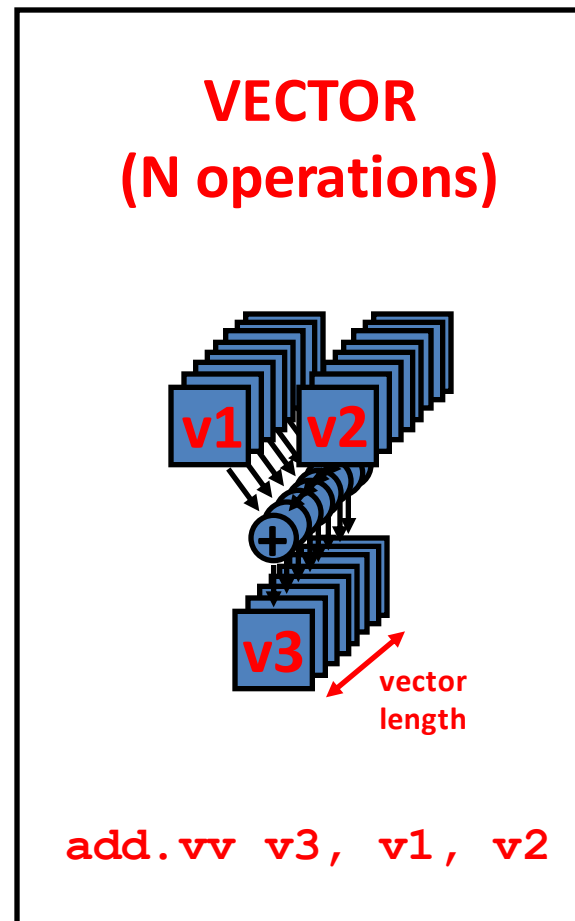
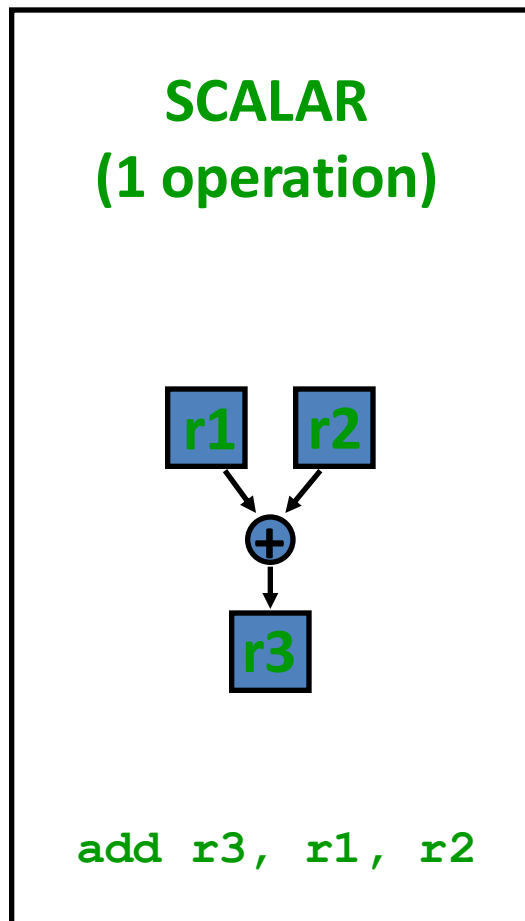


# Vector Processing

# Alternative Model: Vector Processing

- Vector processors have high-level operations that work on linear arrays of numbers: "vectors"



### 3.4.1 Characteristics of Vector Processing

A vector operand contains an ordered set of  $n$  elements, where  $n$  is called the *length* of the vector. Each element in a vector is a scalar quantity, which may be a floating-point number, an integer, a logical value, or a character (byte). Vector instructions can be classified into four primitive types:

$$\begin{aligned} f_1 &: V \rightarrow V \\ f_2 &: V \rightarrow S \\ f_3 &: V \times V \rightarrow V \\ f_4 &: V \times S \rightarrow V \end{aligned} \tag{3.20}$$

where  $V$  and  $S$  denote a vector operand and a scalar operand, respectively. The mappings  $f_1$  and  $f_2$  are unary operations and  $f_3$  and  $f_4$  are binary operations. As

# Vector Instruction

- Given V- a Vector(operand)
  - An ordered set of n elements(n : the *length* of vector)
    - Elements are scalar : floating point number, integer, logical value, character(byte)
- Given S- a scalar (operand)
- Unary Operation- f1 and f2:
  - f1:  $V \longrightarrow V$  example  
 $f_1$           VSQR          Vector square root:  $B(I) \leftarrow \sqrt{A(I)}$

# Vector Instruction contd....

- Unary Operation- f1 and f2:

- f2:  $V \longrightarrow S$  example:

$f_2$       VSUM      Vector summation:       $S = \sum_{I=1}^N A(I)$

- Binary Operation- f3 and f4:

- f3:  $V \times V \longrightarrow V$  example:

$f_3$       VADD      Vector add:       $C(I) = A(I) + B(I)$

- f4:  $V \times S \longrightarrow V$  example:

$f_4$       SADD      Vector-scalar add:       $B(I) = S + A(I)$

**Table 3.5 Some representative vector instructions**

Type	Mnemonic	Description ( $I = 1$ through $N$ )
$f_1$	VSQR	Vector square root: $B(I) \leftarrow \sqrt{A(I)}$
	VSIN	Vector sine: $B(I) \leftarrow \sin(A(I))$
	VCOM	Vector complement: $A(I) \leftarrow \overline{A(I)}$
$f_2$	VSUM	Vector summation: $S = \sum_{I=1}^N A(I)$
	VMAX	Vector maximum: $S = \max_{I=1..N} A(I)$
$f_3$	VADD	Vector add: $C(I) = A(I) + B(I)$
	VMPY	Vector multiply: $C(I) = A(I) * B(I)$
	VAND	Vector and: $C(I) = A(I) \text{ and } B(I)$
	VLAR	Vector larger: $C(I) = \max(A(I), B(I))$
	VTGE	Vector test $>$ : $C(I) = 0$ if $A(I) < B(I)$ $C(I) = 1$ if $A(I) > B(I)$
$f_4$	SADD	Vector-scalar add: $B(I) = S + A(I)$
	SDIV	Vector-scalar divide: $B(I) = A(I)/S$



# VECTOR LENGTH

- vector length of 64.
  - 1. In real world applications vector lengths are not exactly 64.
    - adding just first n elements of a vector ,Vector Length register(VLR) used for this
    - VLR controls the length of any vector operation by defining their length.
    - value cannot be greater than the length of the vector registers. (64 in this case)
  - 2. In real world applications, data in vectors in memory can be greater than the MVL of the processor.
    - we use a technique called Strip Mining



# STRIP MINING

- Splitting data : each vector operation is done for a size less than or equal to MVL.
  - Done by a simple loop with MOD operator as control point.

# STRIP MINING continued...

```
low = 0;
```

```
VL = (n % MVL); /*find odd-size piece using modulo op  
% */
```

```
for (j = 0; j <= (n/MVL); j=j+1)
```

```
{ /*outer loop*/
```

```
for (i = low; i < (low+VL); i=i+1) /*runs for length VL*/
```

```
Y[i] = a * X[i] + Y[i] ; /*main operation* /
```

```
low = low + VL; /*start of next vector*/
```

```
VL = MVL; /*reset the length to MVL*/ }
```