#### Distributed Memory Programming

#### **Distributed Memory Programming**

Wolfgang Schreiner Research Institute for Symbolic Computation (RISC-Linz) Johannes Kepler University, A-4040 Linz, Austria

Wolfgang.Schreiner@risc.uni-linz.ac.at http://www.risc.uni-linz.ac.at/people/schreine

#### **SIMD** Mesh Matrix Multiplication

Single Instruction, Multiple Data

- $\bullet n^2$  processors,
- 3n time.

Algorithm: see slide.

#### **SIMD** Mesh Matrix Multiplication

#### 1. Precondition array

- Shift row i by i-1 elements west,
- Shift column j by j-1 elements north.
- 2. Multiply and add

On processor  $\langle i, j \rangle$ :

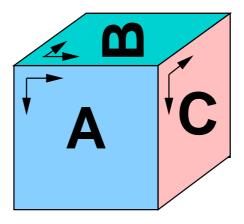
$$c = \sum_{k} a_{ik} * b_{kj}$$

- Inverted dimensions
  - $-\operatorname{\mathsf{Matrix}}\downarrow i, \to j.$
  - $-\operatorname{\mathsf{Processor}}$  array  $\downarrow$  iyproc,  $\rightarrow$  ixproc.
- $\bullet$  *n* shift and *n* arithmetic operations.
- $\bullet n^2$  processors.

Maspar program: see slide.

# SIMD Cube Matrix Multiplication Cube of $d^3$ processors $\int_{y^{0}} \int_{y^{0}} \int_{y^{0$

- $\bullet \mbox{ Map } A(i,j)$  to all P(j,i,k)
- $\bullet \mbox{ Map } B(i,j)$  to all P(i,k,j)



#### **SIMD** Cube Matrix Multiplication

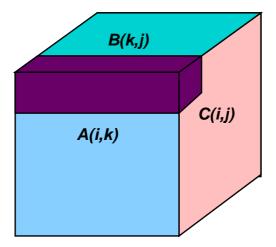
Multiplication and Addition

• Each processor computes single product

$$P_{ijk}: c_{ijk} = a_{ik} * b_{kj}$$

• Bars along x-directions are added

$$P_{0ij}: C_{ij} = \Sigma_k \, c_{ijk}$$



#### **SIMD Cube Matrix Multiplication**

#### Maspar Program

```
int A[N,N], B[N,N], C[N,N];
plural int a, b, c;
a = A[iyproc, ixproc];
b = B[ixproc, izproc];
c = a*b;
for (i = 0; i < N-1; i++)
   if (ixproc > 0)
      c = xnetE[1].c
   else
      c += xnetE[1].c;
if (ixproc == 0) C[iyproc, izproc] = c;
```

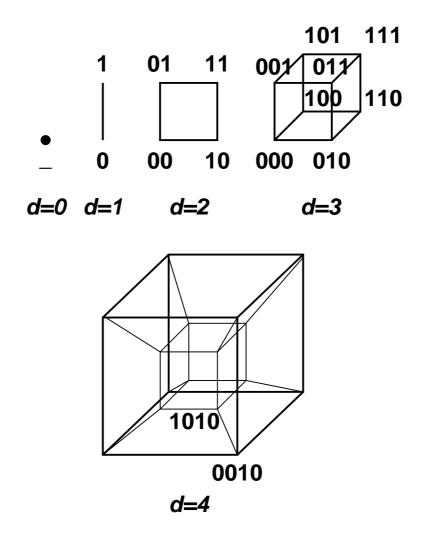
• 
$$O(n^3)$$
 processors,  
•  $O(n)$  time.

### **SIMD Cube Matrix Multiplication**

#### Tree-like summation

```
plural x, d;
...
x = ixproc;
d = 1;
while (d < N) {
    if (x % 2 != 0) break;
    c += xnetE[d].c;
    x /= 2;
    d *= 2;
}
if (ixproc == 0) C[iyproc, izproc] = c;
• O(log n) time
• O(n<sup>3</sup>) processors
Long-distance communication required!
```

#### **SIMD** Hypercube Mat. Multiplication



- d-dimensional hypercube  $\Rightarrow$  processors indexed with d bits.
- $p_1$  and  $p_2$  differ in i bits  $\Rightarrow$  shortest path between  $p_1$  and  $p_2$  has length i.

#### SIMD Hypercube Matrix Multiplication

Mapping of cube with dimension n to hypercube with dimension d.

- Hypercube of  $n^3 = 2^d$  processors  $\Rightarrow d = 3s$  (for some s).
- 64 processors  $\Rightarrow n = 4, d = 6, s = 2.$ Hypercube  $\underline{d_5d_4}$   $\underline{d_3d_2}$   $\underline{d_1d_0}$ Cube x y z
- Embedding algorithm
  - Cube indices in binary form (s bits each)
  - Concatenate indices (3s = d bits)
- Neighbor processors in cube remain neighbors in hypercube.
- Any cube algorithm can be executed with same efficiency on hypercube.

#### SIMD Hypercube Matrix Multiplication

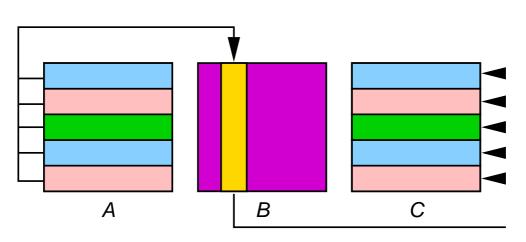
Tree summation in hypercube.

Processors	000	001	010	011	100	101	110	111
Step 1	$r_0$	$s_0$	$r_1$	$s_1$	$r_2$	$s_2$	$r_3$	$s_3$
Step 2	$r_0$		$s_0$		$r_1$		$s_1$	
Step 3	$r_0$				$s_0$			

- Each processor receives value from neighboring processors only.
- Only short-distance communication is required.

*Cube algorithm can be more efficient on hypercube!* 

#### **Row/Column-Oriented Matrix Multi**plication



- 1. Load  $A_i$  on every processor  $P_i$ .
- 2. For all  $P_i$  do:

for j=0 to N-1Receive  $B_j$  from root  $C_{ij} = A_i * B_j$ 

3. Collect  $C_i$ 

Broadcasting of each  $B_j \Rightarrow$  Step 2 takes  $O(N \log N)$  time.

#### **Ring Algorithm**

See Quinn, Figure 7-15.

- Change order of multiplication by
- Using a *ring* of processors.
- 1. Load  $A_i$  and  $B_i$  on every processor  $P_i$ .
- 2. For all  $P_i$  do:

 $p = (i+1) \mod N$  j = ifor k=0 to N-1 do  $C_{ij} = A_i * B_j$   $j = (j+1) \mod N$ Receive  $B_j$  from  $P_p$ 

3. Collect  $C_i$ 

Point-to-point communication  $\Rightarrow$  Step 2 takes O(N) time.

#### Hypercube Algorithm

Problem: How to embed ring into hypercube?

- Simple solution H(i) = i:
  - Ring processor i is mapped to hypercube processor H(i).
  - Massive non-neighbor communication!
- How to preserve neighbor-to-neighbor communication? (see Quinn, Figure 5-13)
- Requirements for H(i):
  - -H must be a 1-to-1 mapping.
  - -H(i) and H(i+1) must differ in 1 bit.
  - -H(0) and H(N-1) must differ in 1 bit.

Can we construct such a function H?

#### **Ring Successor**

Assume H is given.

- Given: hypercube processor number i
- Wanted: "ring successor" S(i)

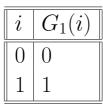
$$S(i) = \begin{cases} 0, & \text{if } i = N-1 \\ H(H^{-1}(i)+1), \text{ otherwise} \end{cases}$$

Same technique for embedding a 2-D mesh into an hypercube (see Quinn, Figure 5-14).

### **Gray Codes**

Recursive construction.

• 1-bit Gray code  $G_1$ 



• n-bit Gray code  $G_n$ 

i	$G_n(i)$	i	$G_n(i)$
0	$0G_{n-1}(0)$	n-1	$1G_{n-1}(0)$
1	$0G_{n-1}(1)$	n-2	$1G_{n-1}(1)$
$\frac{n}{2} - 1$	$0G_{n-1}(\frac{n}{2}-1)$	$\frac{n}{2}$	$1G_{n-1}(\frac{n}{2}-1)$

• Required properties preserved by construction!

$$H(i) = G(i) = i \text{ xor } \frac{i}{2}.$$

#### **Gray Code Computation**

```
C functions.
• Gray-Code
  int G(int i)
  {
    return(i ^ (i/2));
  }
• Inverse Gray-Code
  int G_inv(int i)
  {
     int answer, mask;
    answer = i;
    mask = answer/2;
    while (mask > 0)
     ł
       answer = answer ^ mask;
      mask = mask / 2;
     }
    return(answer);
  }
```

#### **Block-Oriented Algorithm**

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}$$
$$C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix} =$$
$$\begin{pmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{pmatrix}$$

• Use block-oriented distribution introduced for shared memory multiprocessors.

Block-matrix multiplication is analogous to scalar matrix multiplication.

• Use staggering technique introduced for 2D SIMD mesh.

Rotation along rows and columns.

#### • Perform the SIMD matrix multiplication algorithm on whole *submatrices*.

Submatrices are multiplied and shifted.

#### **Analysis of Algorithm**

- $n^2$  matrix, p processors.
  - Row/Column-oriented
    - Computation:  $n^2/p * n/p = n^3/p^2$ .
    - Communication:  $2(\lambda + \beta n^2/p)$
    - -p iterations.

#### • Block-oriented (staggering ignored)

- Computation:  $n^2/p * n/p = n^3/p^2$ .
- Communication:  $4(\lambda + \beta n^2/p)$
- $-\sqrt{p}-1$  iterations.

#### • Comparison

$$\begin{split} & 2p(\lambda+\beta n^2/p) > 4(\sqrt{p}-1)(\lambda+\beta n^2/p) \\ & 2\lambda p + 2\beta n^2 > 4\lambda(\sqrt{p}-1) + 4\beta(\sqrt{p}-1)n^2/p \\ & 1. \ p > 2(\sqrt{p}-1) \\ & 2. \ 1 > 2(\sqrt{p}-1)/p \\ & \text{True for all } p \geq 1. \end{split}$$

## Also including staggering, for larger p the block-oriented algorithm performs better!