
Parallelisation

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Lecture Overview

- Parallelisation for fork/join
- Mapping parallelism to shared memory multi-processors
- Loop distribution and fusion
- Data Partitioning and SPMD parallelism
- Communication, synchronisation and load imbalance.

Approaches to parallelisation

- Two approaches to parallelisation
 - Traditional shared memory. Based on finding parallel loop iterations
 - Distributed memory compilation. Focus on mapping data, computation follows
- Now single address space, physically distributed memory uses a mixture of both.
- Actually, can show equivalence

Loop Parallelisation

- Assume a single address space machine. Each processor sees the same set of addresses. Do not need to know physical location of memory reference.
- Control- orientated approach. Concerned with finding independent iterations of a loop. Then map or schedule these to the processor.
- Aim: find maximum amount of parallelism and minimise synchronisation.
- Secondary aim: improve load imbalance. Inter-processor communication not considered.
- Main memory just part of hierarchy - so use uni-processor approaches.

Loop Parallelisation: Fork/join

- Fork/join assumes that there is a forking of parallel threads at the beginning of a parallel loop
- Each thread executes one or more iterations. Depend on later scheduling policy
- There is a corresponding join or synchronisation at the end
- For this reason loop parallel approaches favour outer loop parallelism
- Can use loop interchange to improve the fork/join overhead.

Loop Parallelisation: Using loop interchange

```
Do i = 1,N
  Do j = 1,M
    a(i+1,j) = a(i,j) +c
  Enddo
Enddo
```

```
Do i = 1,N
  Parallel Do j = 1,M
    a(i+1,j) = a(i,j) +c
  Enddo
Enddo
```

```
Do j = 1,M
  Do i = 1,N
    a(i+1,j) = a(i,j) +c
  Enddo
Enddo
```

```
Parallel Do j = 1,M
  Do i = 1,N
    a(i+1,j) = a(i,j) +c
  Enddo
Enddo
```

Interchange has reduced synchronisation overhead from $O(N)$ to 1.

Parallelisation approach

- Loop distribution eliminates carried dependences and creates opportunity for outer-loop parallelism.
- However increases number of synchronisations needed after each distributed loop.
- Maximal distribution often finds components too small for efficient parallelisation
- Solution: fuse together parallelisable lops.

Loop Fusion

- Fusion is illegal if fusing two loops causes the dependence direction to be changed

```
Do i = 1,N
```

```
    a(i) = b(i) +c
```

```
Enddo
```

```
Do i = 1,N
```

```
    d(i) = a(i+1) +e
```

```
Enddo
```

```
Do i = 1,N
```

```
    a(i) = b(i) +c
```

```
    d(i) = a(i+1) +e
```

```
Enddo
```

- Profitability: Parallel loops should not generally be merged with sequential loops: Tapered fusion

Data Parallelism

- Alternative approach where we focus on mapping data rather than control flow to the machine
- Data is partitioned/distributed across the processors of the machine
- The computation is then mapped to follow the data - typically such that work writes to local data. Local write/owner computes rule.
- All of this is based on the SPMD computational model. Each processor runs one thread executing the same program, operating on the different data
- This means that loop bounds change from processor to processor.

Data Parallelism: Mapping

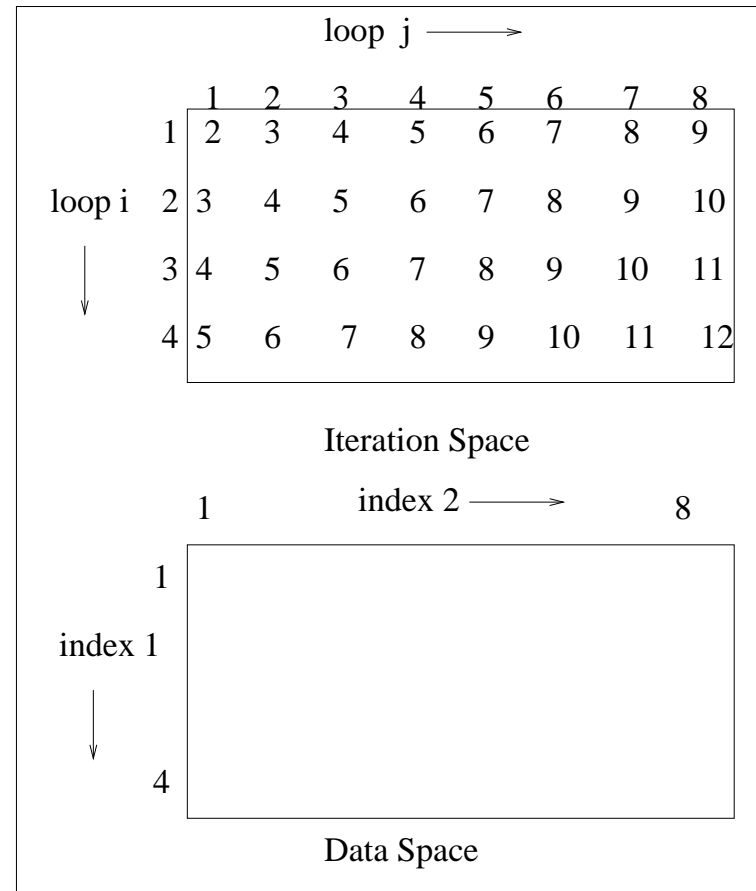
- Placement of work and data on processors. Assume parallelism found in a previous stage
- Typically program parallelism $O(n)$ is much greater than machine parallelism $O(p)$, $n \gg p$
- We have many options as to how to map a parallel program
- Key issue: What is the best mapping that achieves $O(p)$ parallelism but minimises cost
- Costs include communication, load imbalance and synchronisation

Simple Fortran example

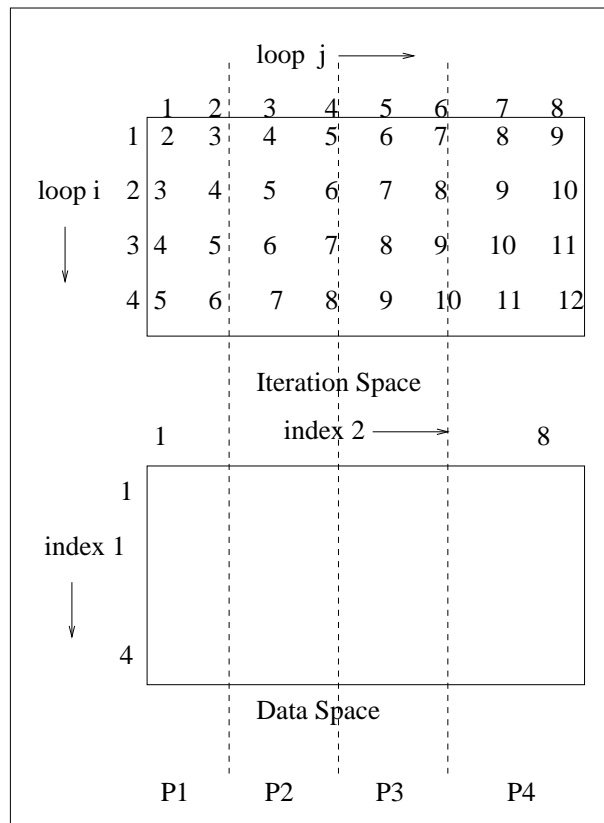
```

Dimension Integer a(4,8)
Do i = 1, 4
  Do j = 1,8
    a(i,j) = i + j
  Enddo
Enddo

```

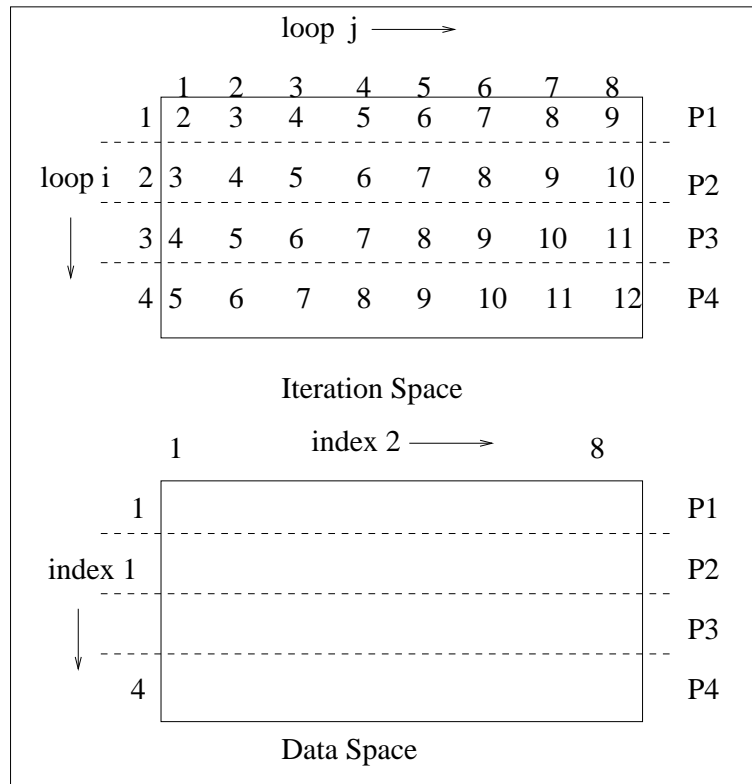


Partitioning by columns of a and hence iterator j : Local writes



```
Dimension Integer a(4,1..2)
Do i = 1, 4           Processor 1
  Do j = 1,2
    a(i,j) = i + j
  Enddo
Enddo
...
Dimension Integer a(4,5..6)
Do i = 1, 4           Processor 3
  Do j = 5,6
    a(i,j) = i + j
  Enddo
Enddo           etc..
```

Partitioning by rows of a and hence iterator i: Local writes



```
Dimension Integer a(1..1,1..8)
Do i = 1, 1           Processor 1
  Do j = 1,8
    a(i,j) = i + j
  Enddo
Enddo
...
Dimension Integer a(3..3,1..8)
Do i = 3, 3         Processor 3
  Do j = 1,8
    a(i,j) = i + j
  Enddo
Enddo           etc..
```

Linear Program representation

```

Do i = 1,16
  Do j = 1,16
    Do k = i,16
      c(i,j) = c(i,j)
              +a(i,k)*b(j,k)
  
```

$$\begin{bmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 1 & 0 & -1 \\ \hline 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} i \\ j \\ k \end{bmatrix} \leq \begin{bmatrix} -1 \\ -1 \\ 0 \\ \hline 16 \\ 16 \\ 16 \end{bmatrix}$$

Polytope $AJ \leq b$. Access matrices $U_c U_a U_b$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}_c \begin{bmatrix} i \\ j \\ k \end{bmatrix}, \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}_a \begin{bmatrix} i \\ j \\ k \end{bmatrix}, \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}_a \begin{bmatrix} i \\ j \\ k \end{bmatrix}$$

Can we automatically generate code for each processor given that writes must be local?

Partitioning: Ex. 1st index: 4 procs: c(16,16), a(16,16),b(16,16)

$$\begin{array}{l}
 \text{Do } i = 5,9 \\
 \quad \text{Do } j = 1,16 \\
 \quad \quad \text{Do } k = i,16 \\
 \quad \quad \quad c(i,j) = c(i,j) \\
 \quad \quad \quad \quad +a(i,k)*b(j,k)
 \end{array}
 \begin{bmatrix}
 -1 & 0 & 0 \\
 0 & -1 & 0 \\
 1 & 0 & -1 \\
 \hline
 1 & 0 & 0 \\
 0 & -1 & 0 \\
 0 & 0 & -1 \\
 \hline
 -1 & 0 & 0 \\
 1 & 0 & 0
 \end{bmatrix}
 \begin{bmatrix}
 i \\
 j \\
 k
 \end{bmatrix}
 \leq
 \begin{bmatrix}
 -1 \\
 -1 \\
 0 \\
 \hline
 16 \\
 16 \\
 16 \\
 \hline
 -5 \\
 9
 \end{bmatrix}$$

Partitioning: Determine local array bounds λ_z, v_z for each processor $1 \leq z \leq p$.

$$\lambda_1 = 1, \lambda_2 = 5, \lambda_3 = 9, \lambda_4 = 13 \quad v_1 = 4, v_2 = 8, v_3 = 12, v_4 = 16$$

Determine local write constraint $\lambda_z \leq U_c \leq v_z, 5 \leq i \leq 9$ and add to polytope

Works for arbitrary loop structures and accesses

Load Balance : 4 procs

```
Do i = 1,16
  Do j = 1,16
    Do k = i,16
      c(i,j) = c(i,j) +a(i,k)*b(j,k)
```

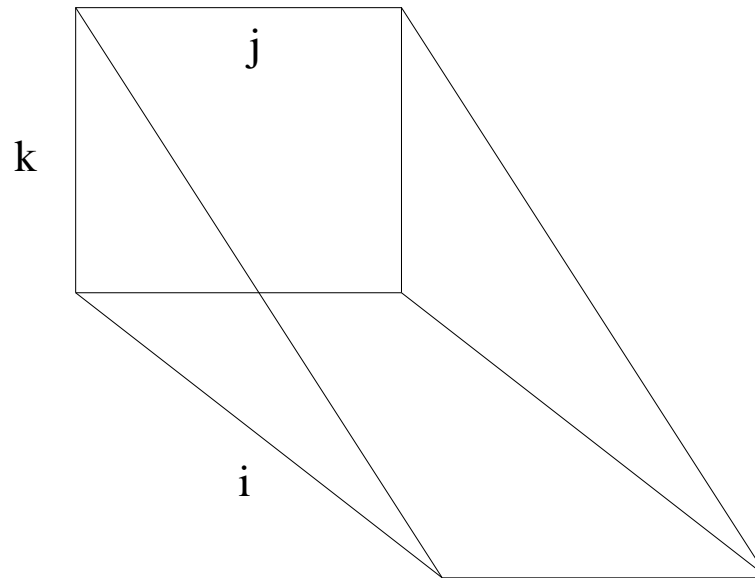
Assuming c, a,b are to be partitioned in a similar manner

How should we partition to minimise load imbalance?

- Row: 928,672,416,160 per processor, load imbalance: 384
- Column: 544 iterations per processor

Why this variation?

Load Balance :



Partition by ""invariant" iterator j.

Can be expressed as a polytope condition

Reducing Communication

We wish to partition work and data to reduce amount of communication or remote accesses

```
Dimension a(n,n) b(n,n)
```

```
Do i = 1,n
```

```
  Do j = 1,n
```

```
    Do k = 1,n
```

```
      a(i,j) = b(i,k)
```

```
    Enddo
```

```
  Enddo
```

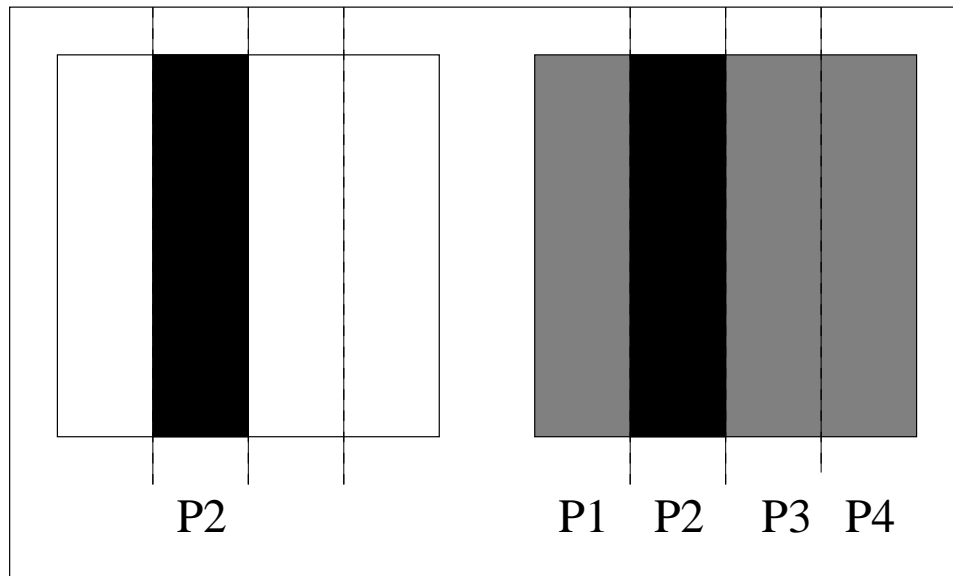
```
Enddo
```

How should we partition to reduce communication?

Reducing Communication :Column Partitioning

Each processor has columns of a and b allocated to it

Look at access pattern of second processor

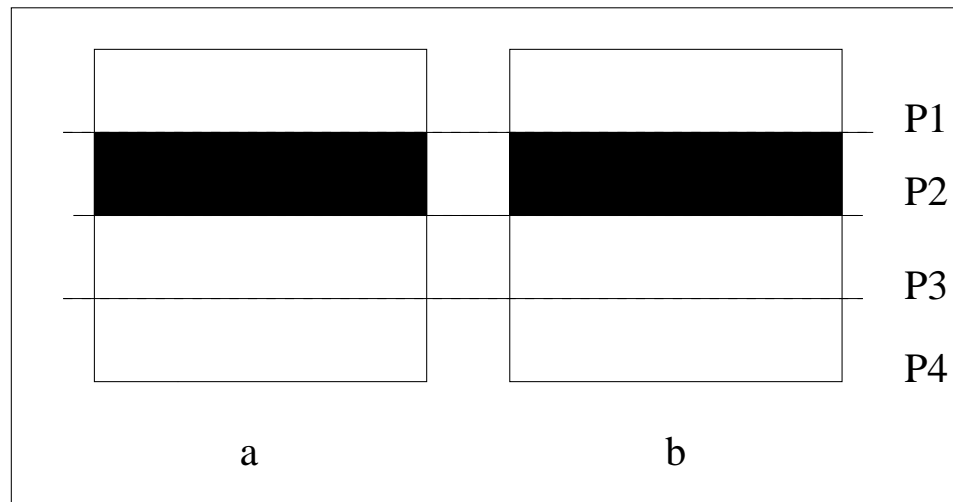


The columns of a scheduled to P2 access all of b $n^2 - \frac{n^2}{p}$ remote access

Reducing Communication :Row Partitioning

Each processor has rows of a and b allocated to it

Look at access pattern of second processor



The rows of a scheduled to P2 access corresponding rows of b .

0 remote accesses.

Alignment

- The first index of a and b have the same subscript $a(i,j)$, $a(i,k)$
- They are said to be aligned on this index
- Partitioning on an aligned index makes all accesses local to that array reference

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}_a, \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}_b$$

Can transform array layout to make arrays more aligned for partitioning.

Find \mathcal{A} such that $\mathcal{A}\mathcal{U}_x$ is maximally aligned with \mathcal{U}_y

Global alignment problem

Synchronisation

- Alignment information can also be used to eliminate synchronisation
- Early work in data parallelisation did not focus on synchronisation
- The placement of message passing synchronous communication between source and sink would (over!) satisfy the synchronisation requirement
- When using data parallel on new single address space machines, have to reconsider this.
- Basic idea, place a barrier synchronisation where there is a cross-processor data dependence.

Synchronisation

```
Do i = 1,16  
  a(i) = b(i)  
Enddo
```

```
Do i = 1,16  
  c(i) = a(i)  
Enddo
```

```
Do i = 1,16  
  a(17-i) = b(i)  
Enddo
```

```
Do i = 1,16  
  c(i) = a(i)  
Enddo
```

- Barrier placed between each loop. But are they necessary?
- Data that is written always local. (localwrite rule)
- Data that is aligned on partitioned index is local.
- No need for barriers here

Summary

- VERY brief overview of auto- parallelism
- Parallelisation for fork/join
- Mapping parallelism to shared memory multi-processors
- Data Partitioning and SPMD parallelism
- Multi-core processor are common place
- Sure to be an active area of research for years to come