## Parallel Algorithm Construction

$\square$ Parallel algorithms for MIMD machines can be divided into 3 categories,
$\square$ these are :
$\square$ Pipelined Algorithms / Algorithmic Parallelism
$\square$ Partitioned Algorithms / Geometric
Parallelism
$\square$ Asynchronous / Relaxed Algorithms

## Pipelined Algorithms

$\square$ Typically each processor forms part of a pipeline and
$\square$ performs only a small part of the algorithm.
$\square$ Data then flows through the system (pipeline ) being operated on by each processor in succession.

## Example

## Pipelined widget assembly machine

$\square$ Say we use a 3 segment pipeline where each of the subtasks ( $A, B$ or $C$ ) is assigned to a segment
$\square$ i.e. the machine is split into 3 smaller machines; one to do step $A$, one for step $B$ and one for step $C$ and which can operate simultaneously.

## Pipelined Algorithms / Algorithmic Parallelism

$\square$ A pipelined algorithm is an ordered set of ( possibly different ) processes in which the output of each process is the input to its successor.
$\square$ The input to the first process is the input to the algorithm
$\square$ The output from the last process is the output of the algorithm.

## Example

$\square$ Say it takes 3 steps $A, B \& C$ to assemble a widget and assume each step takes one unit of time
$\square$ Sequential widget assembly machine:
$\square$ Spends 1 unit of time doing step $A$ followed by 1 unit of time doing step $B$, followed by 1 unit of time doing step $C$
$\square$ So a sequential widget assembler produces 1 widget in 3 time units, 2 in 6 time units etc. i.e. one widget every 3 units

## Example

$\square$ The first machine performs step A on a new widget every time step and
$\square$ passes the partially assembled widget to the second machine which performs step B.
$\square$ This is then passed onto the third machine to perform step $C$

## Example

$\square$ This produces the first widget in 3 time units (as the sequential machine),
$\square$ but after this initial startup time one widget appears every time step.
$\square$ i.e. the second widget appears at time 4 the third widget appears at time 5 etc.


## Pipelined Algorithms

$\square$ In general

- if $L$ is the number of steps to be performed $\square$ and $T$ is the time for each step
$\square$ and n is the number of items (widgets)
$\square$ then Time Sequential $=L T n$
$\square$ and Time Parallel $=[L+n-1] T$


## Geometric Parallelism / Partitioned Algorithms

$\square$ These algorithms arise when there is a natural way to decompose the data set into smaller "chunks" of data,
$\square$ which are then allocated to individual processors.
$\square$ Thus each processor contains more or less the same code but operates on a subset of the total data.
$\square$

## Partitioned Algorithms

$\square$ The solution to these subproblems are then combined to form the complete solution.
$\square$ Depending on the algorithm being solved this combining of solutions usually implies
$\square$ communication synchronization among the processors.
$\square$ Synchronization means constraining a particular ordering of events.

## Partitioned Algorithms

$\square$ To illustrate the difference between pipelined and partitioned algorithms consider the following:
$\square$ Say an algotithm consists of 4 parts $A, B, C$ and $D$ and
$\square$ this algorithm is to operate on a data set E consisting of 4 subsets E1, E2 , E3 and E4
$\square$ (e.g. divide up matrix into submatrix )

## Partitioned Algorithms

$\square$ However in the partitioned algorithm the four processors all perform $A, B, C$ and $D$ but only on a subset of the data


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## Example

> if data needs to be communicated between processors after each iteration of a numerical calculation then this implies synchronization between processes.
> Thus partitioned algorithms are sometimes called synchronous algorithms

## Partitioned Algorithms



## Partitioned Algorithms

$\square$ i.e. In pipelined algorithms the algorithm is distributed among the processors whereas in partitioned algorithms the data is distributed among the processors.

## Example

$\square$ Say we want to calculate $\mathrm{Fi}=\cos \left(\sin \mathrm{e}^{\wedge} \operatorname{sqr}(\mathrm{xi})\right)$ for $\times 1, \times 2$,.....x6 using 4 processors.

## $\square$ Pipelined Version



## Example

$\square$ F1 is produced in 4 time units F2 is produced at time 5
i.e. time $=4+(6-1)=9$ units
$==>$ SPEEDUP $=24 / 9=2.6$


## Example

## Asynchronous / Relaxed Parallelism

$\square$ In relaxed algotithms there is no explicit dependency between processes,
$\square$ as occurs in synchronized algorithms.
$\square$ Instead relaxed algorithms never wait for input.
$\square$ If they are ready they use the most recently available data
$\square$ i.e. time $=8$ units
$==>$ SPEEDUP $=24 / 8=3$
$==>$ EFFICIENCY $=75 \%$
$\square$ Efficiency is calculated by dividing speedup by number of processors
$\square \quad E=S / n$

## Relaxed Parallelism

$\square$ To illustrate this consider the following.
$\square$ Say we have two processors A and B. A produces a sequence of numbers a1, a2 ..
$\square B$ inputs ai and performs some calculation $F$ which uses ai.
$\square$ Say that $B$ runs much faster than $A$.

## Example

## $\square$ Synchronous Operation

$\square$ A produces al passes it to $B$ which calculates F1;
$\square$ A produces a2 passes it to $B$ which calculates F2;
$\square$ i.e. B waits for A to finish (since B is faster than A ) etc..

## Relaxed Parallelism

$\square$ The idea in using asynchronous algorithms is that all processors are kept busy and never remain idle (unlike synchronous algorithms) so speedup is maximized.
$\square$ A drawback is that they are difficult to analyse ( because we do not know what data is being used ) and
$\square$ also an algorithm that is known to work (e.g. converge) in synchronous mode may not work (e.g diverge) in asynchronous mode.

## Relaxed Parallelism

$\square$ Say we have 3 processors
$\square$ P1 : given $x, P 1$ calculates $F(x)$ in time $\dagger 1$, units and sends it to P3
$\square \mathrm{P} 2$ : given $\mathrm{y}, \mathrm{P} 2$ calculates $\mathrm{F}^{\prime}(\mathrm{y})$ in time t 2 units and sends it to P 3
$\square$ P3 : given $a, b, c, P 3$ calculates $d=a-b / c$ in time t3 units;
$\square$ if $|d-a|>$ Epsilon then $d$ is sent to P1 and P2 otherwise $d$ is output.

## Example

## $\square$ Asynchronous Operation

$\square$ A produces al passes it to $B$ which calculates $F 1$
$\square$ but now $A$ is still in the process of computing a2
$\square$ so instead of waiting B carries on and calculates F2 (based on old data i.e. al and therefore may not be the same as F2 above ) and
$\square$ continues to calculate $F$ using the old data until a new input arrives
$\square$ e.g. Fnew $=$ Fold $+a i$

## Relaxed Parallelism

$\square$ Consider the Newton Raphson iteration for solving
$\square F(x)=0$
$\square$ where $F$ is some non-linear function
$\square$ i.e. $X n+1=X n-F(X n) / F^{\prime}(X n)$......(1)
generates a sequence of approximations to the root, starting from a value $X 0$.

## Example

## $\square$ Serial Mode

$\square \mathrm{P} 1$ computes $\mathrm{F}(\mathrm{Xn})$
$\square$ then P2 computes $F^{\prime}(X n)$
$\square$ then P3 computes $\mathrm{Xn}+1$ using (1)
$\square$ So time per iteration is $\dagger 1+\dagger 2+\dagger 3$

- If $k$ iterations are necessary for convergence then total time is $k(t 1+t 2+t 3)$

| Example |
| :---: |
| $\square$ Synchronous Parallel Mode. <br> $\square \mathrm{P} 1$ and P 2 compute $\mathrm{F}\left(\mathrm{Xn}_{n}\right)$ and $\mathrm{F}^{\prime}\left(\mathrm{Xn}_{n}\right)$ simultaneously and $\square$ when $B O T H$ have finished the values $F\left(X_{n}\right)$ and $F^{\prime}\left(X_{n}\right)$ are used by P 3 to compute $\mathrm{Xn}+1$ <br> $\square$ Time per iteration is $\max (+1,+2)++3$ <br> $\square$ Again $k$ iterations will be necessary so total time is $\mathbf{k}$ $\begin{aligned} & {[\max (+1,+2)++3]} \\ & X 1=X 0-F(X O) / F^{\prime}(X 0) \ldots \text {...etc } \end{aligned}$ |
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## Relaxed Parallelism

## $\square$ Asynchronous Parallel Mode

$\square \mathrm{P} 1$ and P 2 begin computing as soon as a new input value is made available by $P 3$ and they are ready to receive it,
$\square \mathrm{P} 3$ computes a new value using (1) as soon as EITHER P1 OR P2 provide a new input
$\square$ i.e. ( 1 ) is now of the form
$\square \mathbf{X n + 1}=\mathbf{X n}-\mathbf{F}(\mathbf{S X i}) / \mathrm{F}^{\prime}(\mathbf{X i})$

