## Introduction

- Sample data for program will be on the web today
- Reading
- Today OpenMP \& HPF
- Thursday DSM papers
- one paper is only available from the library


## OpenMP

- Support Parallelism for SMPs
- provide a simple portable model
- allows both shared and private data
- provides parallel do loops
- Includes
- automatic support for fork/join parallelism
- reduction variables
- atomic statement
- one processes executes at a time
- single statement
- only one process runs this code (first thread to reach it)


## Sample Code

```
program compute_pi
    integer n , i
    double precision \(\mathrm{w}, \mathrm{x}\), sum, pi, f, a
c function to integrate
    \(f(a)=4 . d 0 /\left(1 . d 0+a^{*} a\right)\)
    print *, \021Enter number of intervals: \021
    read *,n
c calculate the interval size
    \(\mathrm{w}=1.0 \mathrm{~d} 0 / \mathrm{n}\)
    sum \(=0.0 \mathrm{~d} 0\)
!\$OMP PARALLEL DO PRIVATE(x), SHARED(w)
!\$OMP\& REDUCTION(+: sum)
    do \(i=1, n\)
            \(\mathrm{x}=\mathrm{w}\) * \((\mathrm{i}-0.5 \mathrm{~d} 0)\)
            sum \(=\operatorname{sum}+f(x)\)
    enddo
    \(\mathrm{pi}=\mathrm{w}\) * sum
    print *, \021computed pi = \021, pi
    stop
    end
```


## HPF Model of Computation

- goal is to generate loosely synchronous program
- original target was distributed memory machines
- Explicit identification of parallel work
- forall statement
- Extensions to FORTRAN
- the forall statement has been added to the language
- the rest of the HPF features are comments
- any HPF program can be compiled serially
- Key Feature: Data Distribution
- how should data be allocated to nodes?
- critical questions for distributed memory machines
- turns out to be useful for SMP too since it defines locality


## HPF Language Concepts

- Virtual processor
- an abstraction of a CPU
- can have one and two dimensional arrays of VPs
- each VP may map to a physical processor
- several VP's may map to the same processor
- Template
- a virtual array (no data)
- used to describe how real array are aligned with each other
- templates are distributed onto to virtual processors
- Align directives
- expresses how data different arrays should be aligned
- uses affine functions
- align element I of array A with element I+3 of B


## Distribution Options

- BLOCK
- divide data into N (one per VP) contiguous units
- CYCLIC
- assign data in round robin fashion to each processor
- BLOCK(n)
- groups of $n$ units of data are assigned to each processor
- must be exactly (array size)/n virtual processors
- CYCLIC(n)
- n units of contiguous data are assigned round robin
- CYCLIC is the same as CYCLIC(1)


## Computation

- Where should the computation be performed?
- Goals:
- do the computation near the data
- non-local data requires communication
- keep it simple
- HPF compilers are already complex
- Compromise: "owner computes"
- computation is done on the node that contains the rhs of a statement
- non-local data for the lhs operands are send the node as needed


## Finding the Data to Use

- Easy Case
- the location of the data is known at compile time
- Challenging case
- the location of the data is a known (invertable) function of input parameters such as array size
- Difficult Case (irregular computation)
- data location is a function of data
- indirect array used to access data $A[$ index $[I], j]=$...


## Challenging Case

- Each processor can identify its data to send/recv
- use a pre-processing loop to identify the data to to move
for each local element I
receive_list = global_to_proc(f(I))
send_list = global_to_proc( $\left.f^{-1}(\mathrm{I})\right)$
send data in send_list and receive data in receive_list
for each local rhs element I
perform the computation


## Irregular Computation

- Pre-processing step requires data to be sent
- since we might need to access non-local index arrays
- two possible cases
- gather $\mathrm{a}(\mathrm{I})=\mathrm{b}(\mathrm{u}(\mathrm{I}))$
- pre-processing builds a receive list for each processor
- send list is known based on data layout
- scatter $\mathrm{a}(\mathrm{u}(\mathrm{I}))=\mathrm{b}(\mathrm{I})$
- pre-processing builds a send list for each processor
- receive list is known based on data layout


## Communication Library

- How is it different from pvm?
- abstraction based on distributed, but global arrays
- provides some support for index translation
- pvm has local arrays
- multicast is in one dimension of a array only
- shifts and concatenation provided
- special ops for moving vectors of send/recv lists
- precomp_read
- postcomp_write
- Goals
- written in terms of native message passing
- tries to provide a single portable abstraction to compile to


## Performance Results

- How good are the speedup results?
- only one application shown
- speedup is similar to hand tuned message passing program
- one extra $\log (\mathrm{n})$ communication operations slows perf
- how good is the hand tuned program?
- speedup is only 6 on 16 processors
- What is figure 4 showing?
- compares performance on two different machines
- no explanation
- is this showing the brand $x$ is better then brand $y$ ?
- does it show that their compiler doesn't work on brand $y$ ?
- lesson: figures should always tell a story
- don't require the reader to guess the story


## Communitivity Analysis:Target Environment

- Shared memory multi-processors
- Object oriented programs
- $\mathrm{C}^{++}$class methods
- pointer based graph data structures
- Sources of parallelism
- method invocation
- methods may be invoked
- recursively
- simple looping constructs (converted to tail recursion)


## Analysis

- Determine if two method invocations commute
- intuitive definition: can be performed in any order
- a followed by $b(a ; b)$ is the same as $b$ then $a(b ; a)$
- Technique
- symbolic evaluation
- generate symbolic results of running $a ; b$ and $b ; a$
- like running a method but expressions not data
- compare two results
- invar analysis - are the variables the same?
- Need to know basic commutative ops (e.g. addition)
- sub-method invocation
- are multi-sets of different invocations the same


## Performance Issues

- Method Size
- methods should be the "natural" size
- too small - not enough work for overhead
- too largew -results in a load imbalance
- Synchronization
- need to provide mutex over shared data
- granularity an important parameter
- too small - lock overhead dominates
- too large - reduce potential parallelism
- Compiler can change granularity
- start with one lock per method invocation
- user lock "coarsening" to merge locks across invocations


## Lock Granularity

- Hard to know correct lock size at compile time Solution: use runtime adaptation
- Generate multiple versions of methods
- each uses a different lock granularity
- provide a way to switch between version
- Adaptation
- run one at a time and gather timing data for each one
- select best one
- need to make sure samples are representative


## Questions About the Technique

- Are the speedups good?
- $50 \%$ is not bad for an automatic tool
- Is the technique general?
- Has only tried two programs
- these were the target applications from the start
- works for recursive graph structures
- how big is this application domain?
- Will it work and play with other approaches?
- Can data parallelism be used for part of the code?

