Towards High-Level Programming Support for Scientific Computing on Clusters

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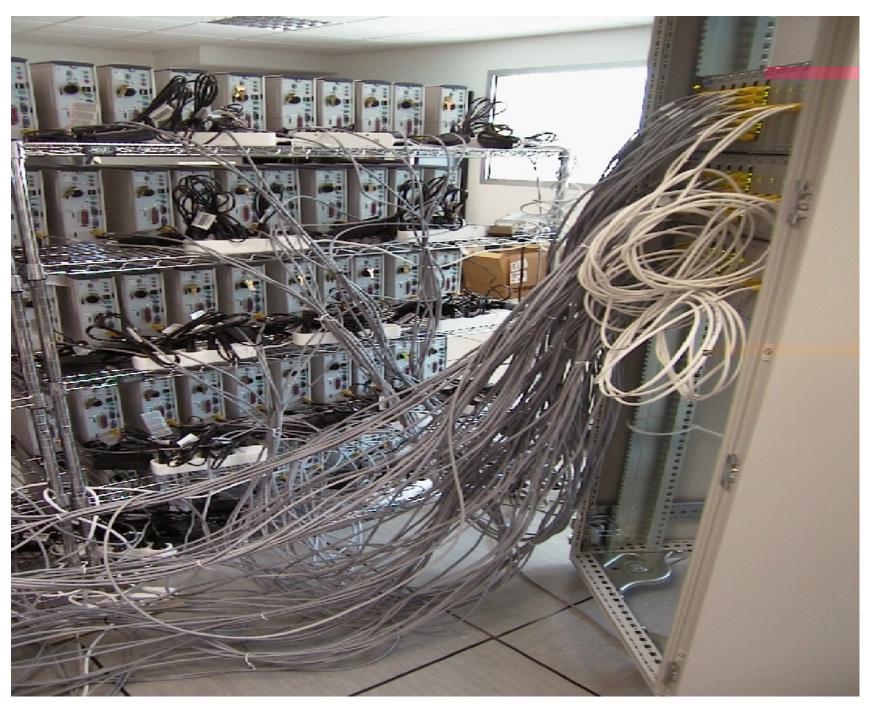
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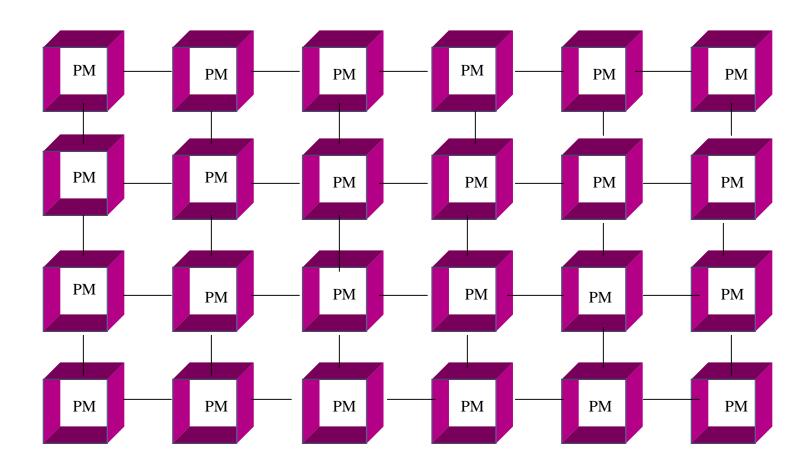
OUTLINE

- 1 Introduction
- 2 Distribution and Alignment Control
- 3 Irregular and Dynamic Problems
- 4 Generalizing the Concept of Distribution
- 5 Future Aspects
- 6 Conclusion



Source: i-Cluster, University of Grenoble, France

Abstract Cluster Architecture



Programming Models

- What is the right model for parallel programming of clusters? -- A viable compromise must be found between the conflicting goals of
 - expressivity
 - portability
 - performance
- For performance-oriented programming, the message-passing model has been traditionally adopted.

This Talk ...

- discusses high-level programming support for dataintensive scientific parallel programming of clusters that can provide sufficient performance, including irregular and dynamic problems
- focuses on high-level specification of
 - distribution of data and work
 - affinity relationships
 - communication

The Problem: Running a single large-scale scientific application in parallel on a cluster. Can the application programmer be provided with higher-level support than message passing without inacceptable loss of performance?

Initial Assumptions

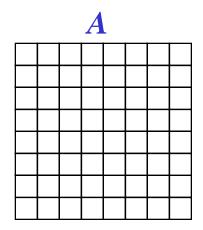
- Homogeneous cluster
- Single-processor nodes
- Each node operates in a separate address space
- Abstract network topology

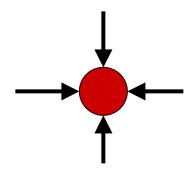
MPI vs. HPF

An Almost Trivial Example (Jacobi)

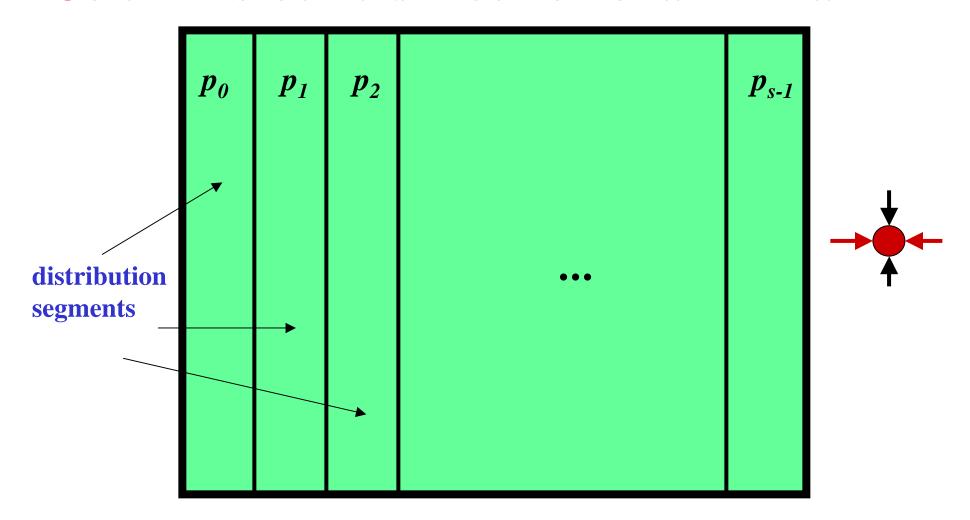
Sequential Jacobi Kernel

```
do while (.not. converged)
  do J=1,N
     do I=1,N
         B(I,J) = 0.25 * (A(I-1,J)+A(I+1,J)+
                        A(I,J-1)+A(I,J+1)
     end do
   end do
  A(1:N,1:N) = B(1:N,1:N)
end do
```

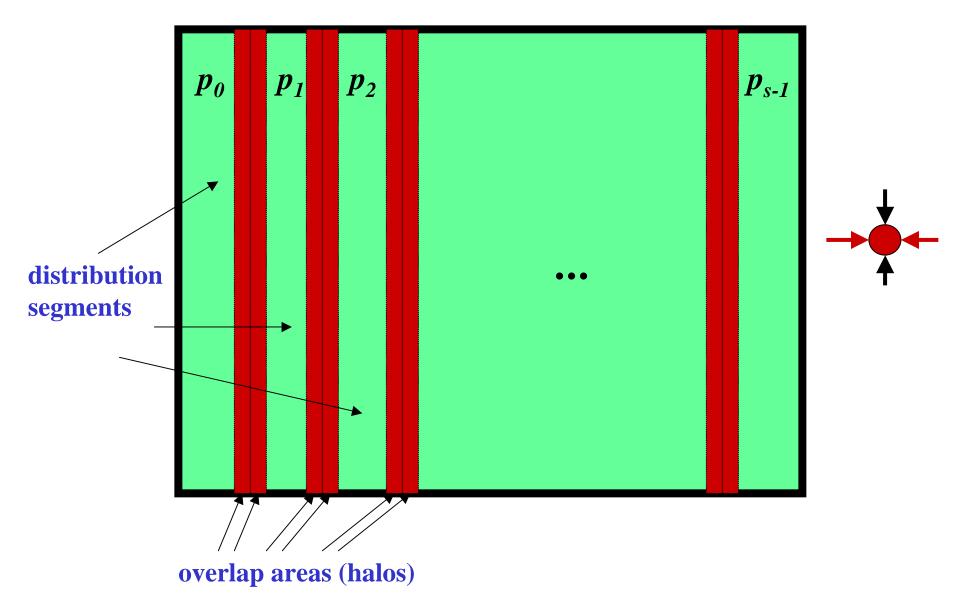




Column-block distribution of a 2D-matrix



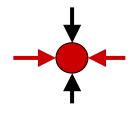
Column-block distribution of a 2D-matrix



Parallel Jacobi with MPI

initialize MPI

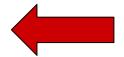
```
do while (.not. converged) do J=1,M computation B(I,J)=0.25*(A(I-1,J)+A(I+1,J)+A(I,J+1)) end do end do A(1:N,1:N)=B(1:N,1:N)
```



```
 if \ (MOD(myrank,2) . eq. \ 1) \ then \ call \ MPI\_SEND(B(1,1),N,...,myrank-1,..) \\  call \ MPI\_RCV(A(1,0),N,...,myrank-1,..) \\  if \ (myrank . lt. \ s-1 \ then \\  call \ MPI\_SEND(B(1,M),N,...,myrank+1,..) \\  call \ MPI\_RCV(A(1,M+1),N,...,myrank+1,..) \\  end if \\  else \ ... \\  ... \\  ... \\
```

Parallel Jacobi with HPF

processors *P(NUMBER_OF_PROCESSORS)* distribute(*,BLOCK) onto *P* :: *A*, *B*



```
do while (.not. converged) global do J=1,N computation B(I,J) = 0.25 * (A(I-1,J)+A(I+1,J)+A(I,J+1)) end do end do A(I:N,I:N) = B(I:N,1:N)
```

Communication is automatically generated by the compiler

Observations

- The HPF code is far simpler than the MPI code
- The compilers can generate from that HPF code an

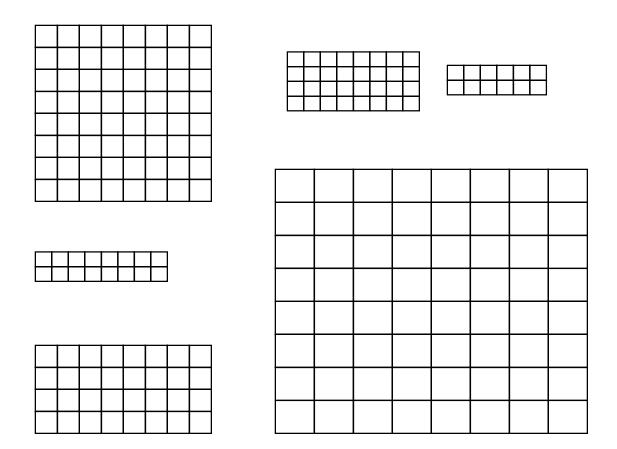
The Problem: Irregular, Dynamic, and Adaptive Aplications

- Examples: applications working with semistructured grid collections, unstructured grid codes, spectral transform codes
- Typical features: Irregular and/or dynamically varying data structures, data access and dependence patterns, data distribution, work patterns and work distributions

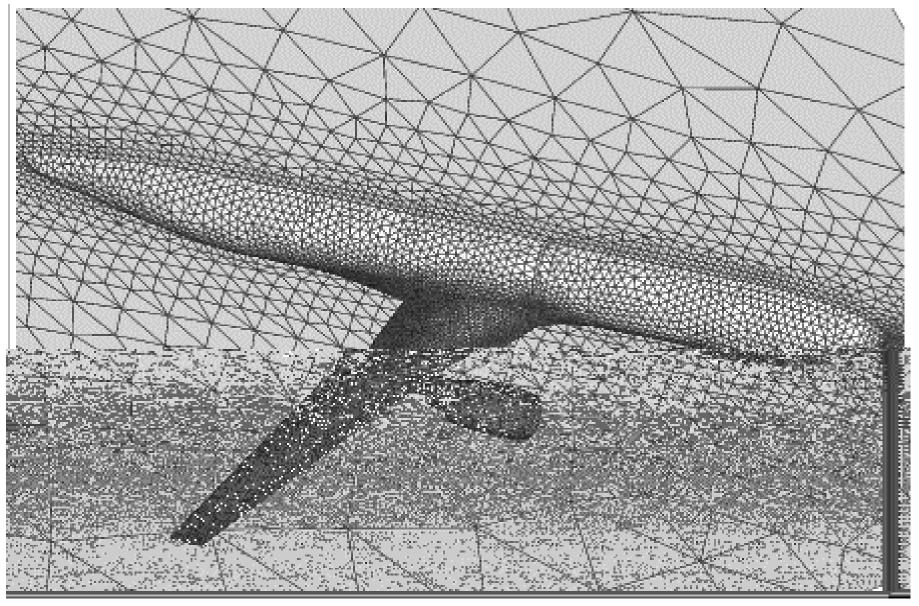
Semi-structured Grid Collections

- Irregularly structured sets of structured grids that can be processed in parallel:
 - multiblock
 - parallel multigrid
 - structured AMR (SAMR)
- In order to exploit two-level parallelism and achieve load balancing, it is necessary to
 - distribute grids to subsets of processors
 - allow irregular distributions

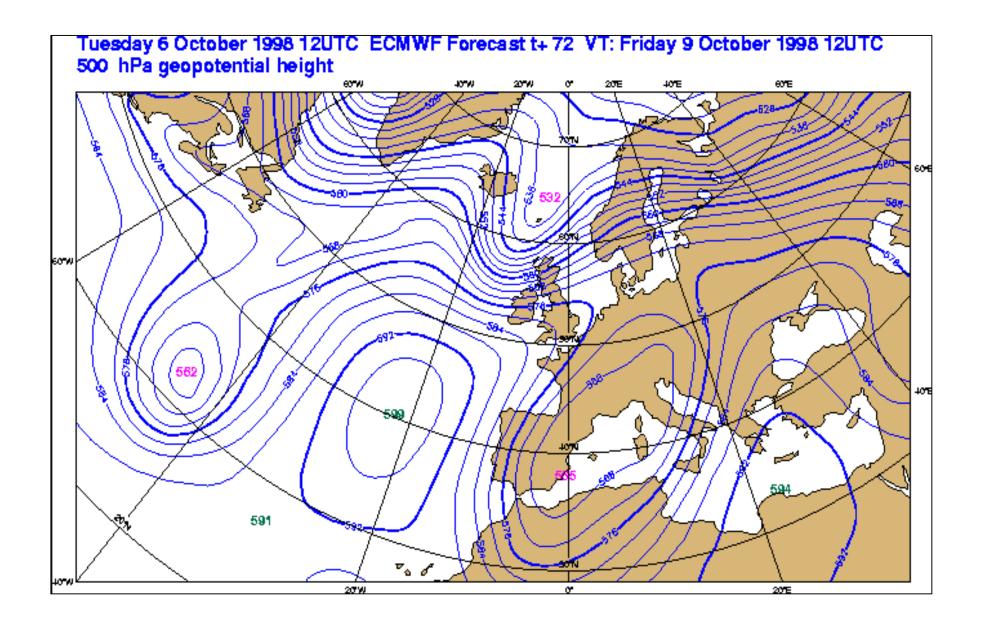
Example: A Multiblock Grid Collection



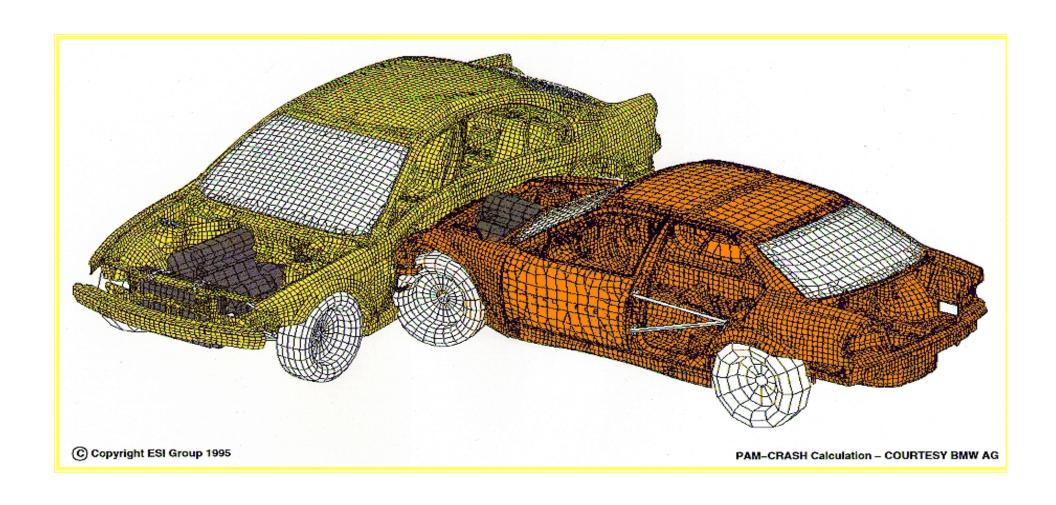
An Unstructured Grid



Source: Dimitri Mavriplis, ICASE, NASA Langley Research Center

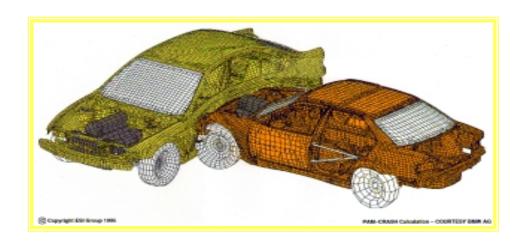


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Finite Element Code for Crash Simulation



Access to unstructured meshes requires at least 2 levels of indirection.

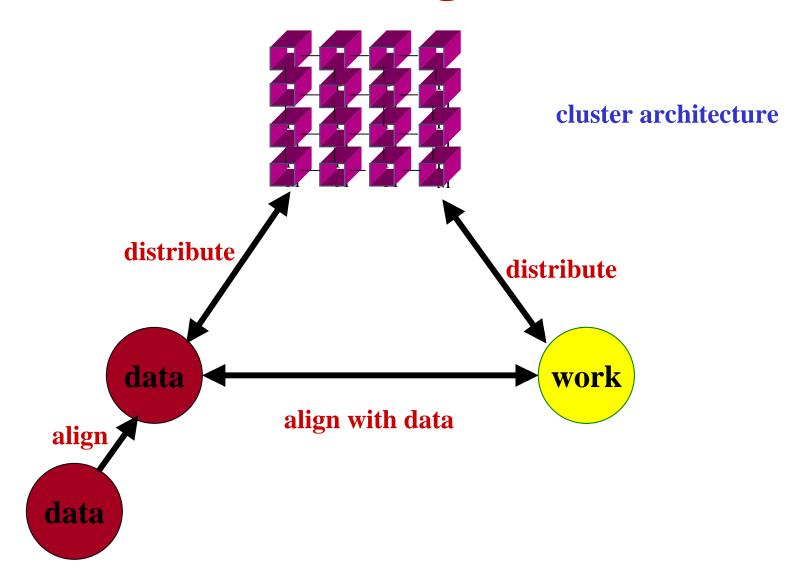
```
REAL :: X(3,N_NODES), F(6,N_NODES),
INTEGER :: IX(4,N_ELEMS) !mesh connectivity
...
do i = 1, N_ELEMS
    do i = 1, 4
        F(:,IX(K,I)) = ...+F(:,IX(K,I))+ ...
    end do
end do
```

access patterns cannot be analyzed at compile-time

Why Simple Distribution Strategies are not Adequate for Irregular Problems

- Regular data distributions may not reflect locality in physical space
- Regular data distributions may not support load balancing for irregularly distributed objects
- the owner-computes paradigm does not allow expression of affinity between data and work distribution
- static distribution strategies cannot reflect dynamic changes of data and computation structures

Distribution and Alignment Control



Requirements for a More General Distribution Model

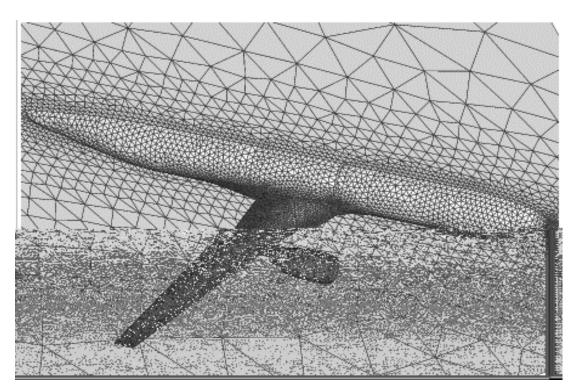
- Generalize mappings index space ---> processors
- Allow distributions to subsets of processors
- Generalize affinity specifications
 - data alignment
 - work / data alignment
- Allow distributions and alignments to change dynamically
- Improved (high-level) control of communication

Data Distribution for Unstructured Grids

Problem: index space locality does not reflect locality in 3D space

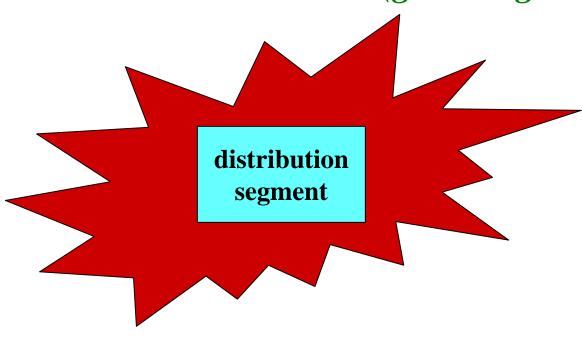
Solution 1: *indirect distribution*

Solution 2: general block distribution combined with reordering of elements

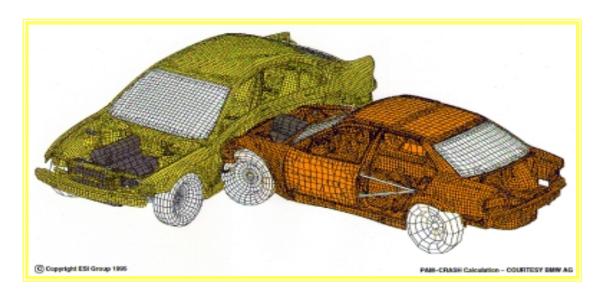


Two Methods for Communication Control

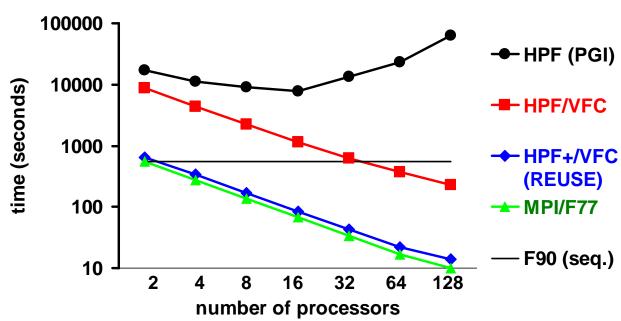
- Schedule Reuse Control: make communication schedules explicitly accessible objects and control their evaluation and reuse
- Halo management: allow explicit control of irregular communication halos (ghost regions)



The Effect of Schedule Reuse and Halo Management



FEM crash simulation kernel



Language Extensions

(HPF/Earth Simulator)

- ☐ Schedule Reuse
- ☐ Halo Control
- ☐ Purest proc's



Example: Sparse Matrix Vector Product

- Generate the matrix in **distributed sparse format**
 - sparse representation (e.g., compressed row storage)
 - data distribution across memories
- For each distribution segment, execute a separate thread computing a (parameterized) local matrix-vector product
- Combine the local vectors determined by the threads (global reduction)

MRD/CRS Sparse Matrix Distribution

0 53	0 0	0	0 0 0
0 0	0 0	0	0 21 0
19 0	0 0	0	0 0 16
0 0	0 0	0	72 0 0
0 0	0 17	0	0 0 0
0 0	0 0	93	0 0 0
0 0	0 0	0	0 13 0
0 0	0 0	44	0 0 19
0 23	69 0	37	0 0 0
27 0	0 11	0	0 64 0

\mathbf{D}_0	C_0	R ⁰
53 19 17 93	2 1 4 5	1 2 2 3 3
•		3 4 5 5

\mathbf{D}^1	C^1	R ¹
21 16 72 13	$\begin{bmatrix} 2\\3\\1\\2 \end{bmatrix}$	1 1 2 3 4
•		4 4 5

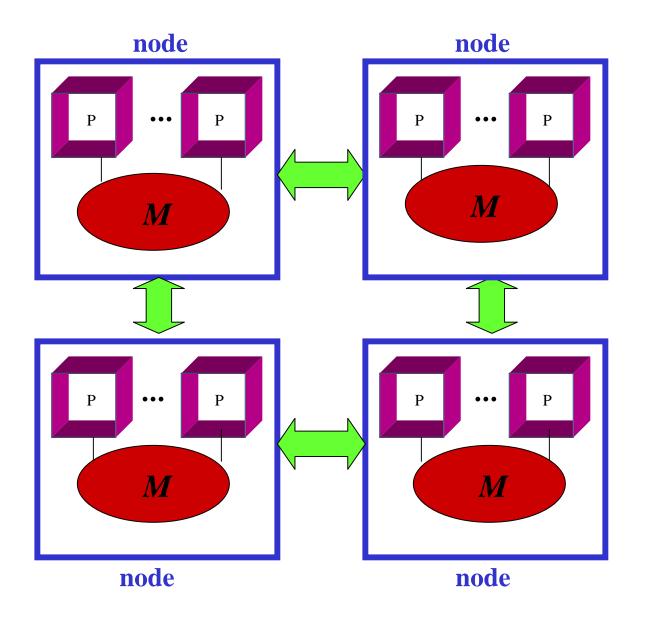
D ² 23 69 27	$\begin{bmatrix} \mathbf{C}^2 \\ 2 \\ 3 \\ 1 \\ 4 \end{bmatrix}$	$ \begin{array}{c} \mathbf{R}^2 \\ \hline 1 \\ 1 \\ 3 \end{array} $
11	4	5

44 1 1 19 4 3 37 1 4 64 3 5	D ³	C ³	R ³
	19 37	4	3

Distributed Sparse Matrix-Vector

```
integer :: NP = number_of_processors()
processors :: P(NP)
real, sparse(CRS(D,C,R,q,L1,U1,L2,U2,...)) :: A(N,M)
method mat\_vec\_loc(u,D,C,R)
real :: D(q(u)); integer I,K,C(q(u)),R(L1(u):U1(u)+1)
\mathbf{do}\ I = L1(u), U1(u)
  TS(u,1:N) = 0.0
  do K = R(I), R(I+1)-1
      TS(u,I) = TS(u,I) + D(K) * B(C(K)+L2(u))
   end do
end do
end mat vec loc
do independent u=1:NM, on home (A(L1(u):U1(u),L2(u):U2(u)))
   call mat vec loc(u,D(u,:),C(u,:),R(u,:))
```

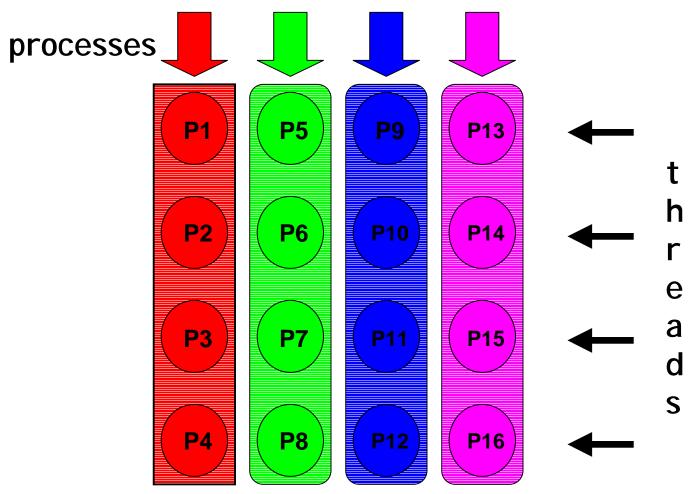
SMP Clusters



Programming Approaches to SMP Clusters

- MPI only
- OpenMP with HPF-like distribution directives
- HPF with OpenMP-extrinsics option
- HPF-like approach based on topology specification

Specification of Cluster Topology



```
processors r(4,4)
nodes n(4)
distribute r(*,block)onto n
```

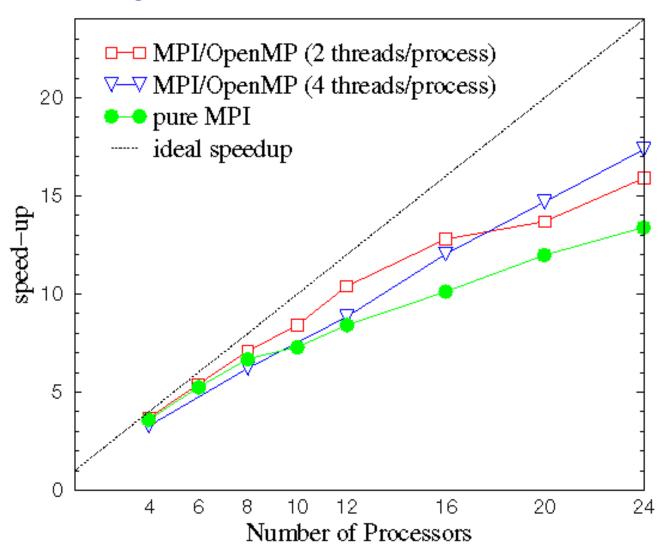
Example:

```
processors r(4,4)
nodes n(4)
distribute r(*,block) onto n

real A(N,M)
distribute(block, block) onto r :: A
```

VFC: Pure MPI vs. OpenMP/MPI

3D Medical Image Reconstruction Kernel on a 6x4 PC Cluster

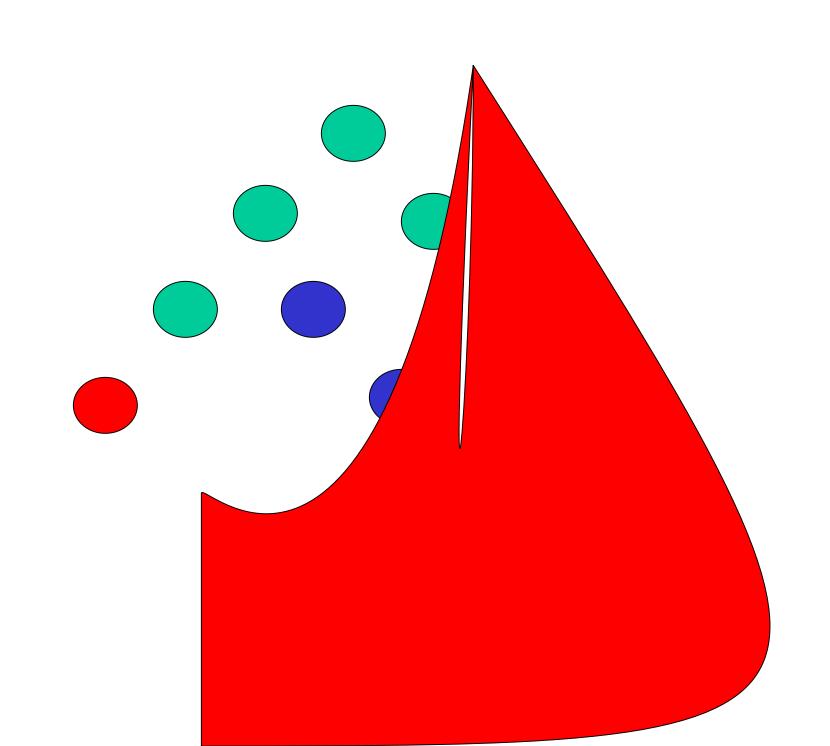


Future Aspects

- Generalizing distribution in an object-based framework
- Trends in compiling
- (Semi-)automatic performance tuning

Generalized Data Distributions

- Distribution of arbitrary data structures
- User-defined specifications
- Distributions as first-class objects with methods
- Affinity control



Distributions as Objects I

- **Distribution object:** a mapping
- $\delta: \mathbf{I} \longrightarrow \mathbf{P}(\mathbf{J})$

- I: data structure index domain
- J: processor index domain
- **Distribution type:** a parameterized specification $\Delta(I,J,...)$, where I and J represent index domains. An *instantiation*, parameterized by an array index domain, I^A , and a processor index domain, I^P , yields a distribution object $\delta = \Delta(I^A, I^P,...)$.
- Distribution types may be *intrinsic* or *user-defined*.

Example: a General Block Distribution

real A(N1,N2)

processors :: P(8)

distribute(*, H) :: A

$$\delta(*,1:h2-1) = \{p1\}$$

•••

distribution

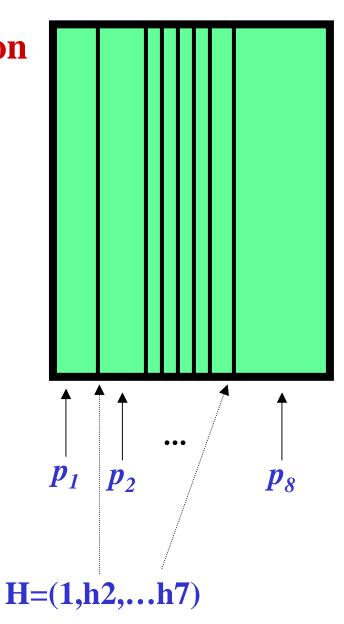
$$\delta(*, h7:N2) = \{p8\}$$

$$\lambda(p1) = [1:N1,1:h2-1]$$

•••

segments

$$\lambda(p8) = [1:N1, h7:N2]$$



Hans P. Zima
Cluster2001, Newport Beach,
California

Example: User-defined Distribution TypeIndirect Distribution

```
dist_type indirect(map)
array A(:)
processors R(:)
integer map(:)
for every I=1, size(A)
     map A(I) to R(map(I))
endfor
```

Distributions as Objects II

- Intrinsic methods for distribution objects:
 - **DISTRIBUTE**(δ ,A,P) generate a distributed data structure
 - **OWNER**(δ ,A,P,i) determine the owner of A(i)
 - **SEGMENT**(δ ,A,P,p) determine the distribution segment of processor p
 - **REDISTRIBUTE**(δ ,A,P, δ ') redistribute based on δ '
 - INC_REDISTRIBUTE(δ ,A,P,inc) redistribute incrementally based on inc

- ...

• Alignment can be understood in this framework as a class of special distribution constructors

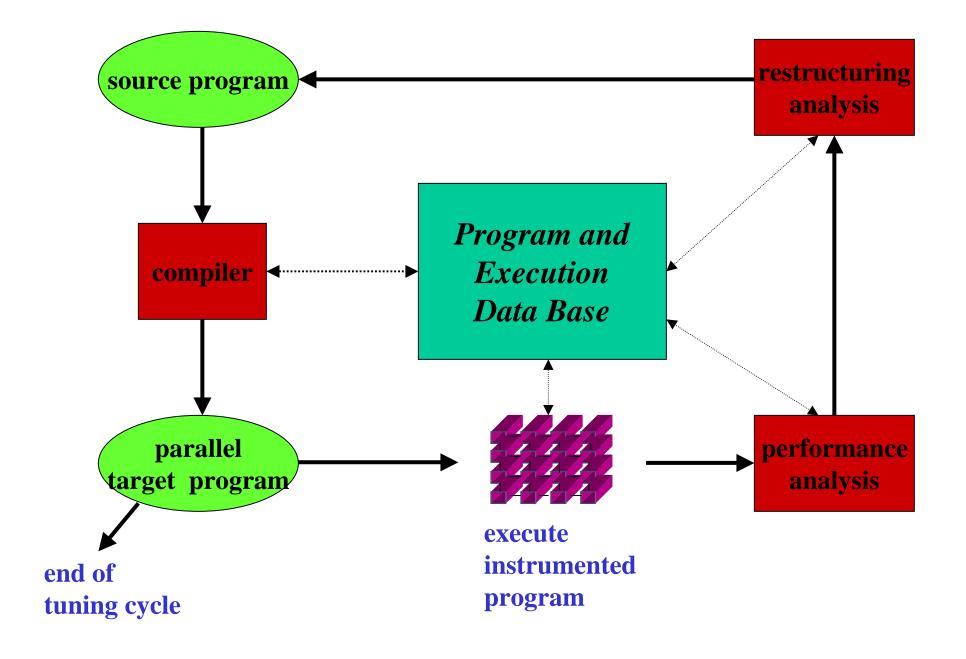
Some Trends in Compiling

- runtime compilation: interaction compiler-runtime system
- feedback-oriented compilation: interaction of compiler with dynamic performance analysis subsystem
- self-adapting software ---> Jack Dongarra's talk
 - Atlas
 - PhiPac
 - FFTW

- ...

• intelligent analysis and restructuring

Iterative Performance Tuning



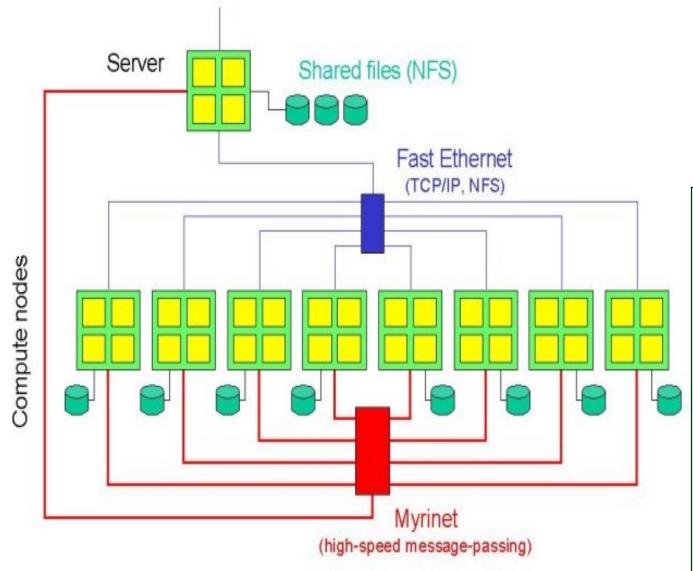
European Cluster Activities

A Few Examples

- ZAMpano Cluster, Juelich, Germany, 8x4 PIII
- Parnass2 Cluster, Bonn, Germany, 72x2 PII
- Gravitor II Cluster, Geneva, Switzerland, 132 PII/PIII
- i-Cluster, Grenoble, France, 216 PIII
- PC2 Cluster, Paderborn, Germany http://www.upb.de/pc2
- SARA Beowulf Cluster, Amsterdam #63 http://www.sara.nl/beowulf
- SUN HPC 4500 400MHz Cluster, Defense, Stockholm, 896 p,#57
- CLIC PIII Cluster, Chemnitz, Germany 528p, #156

http://www.tu-chemnitz.de/urz/anwendungen/CLIC

Juelich ZAMpano Cluster



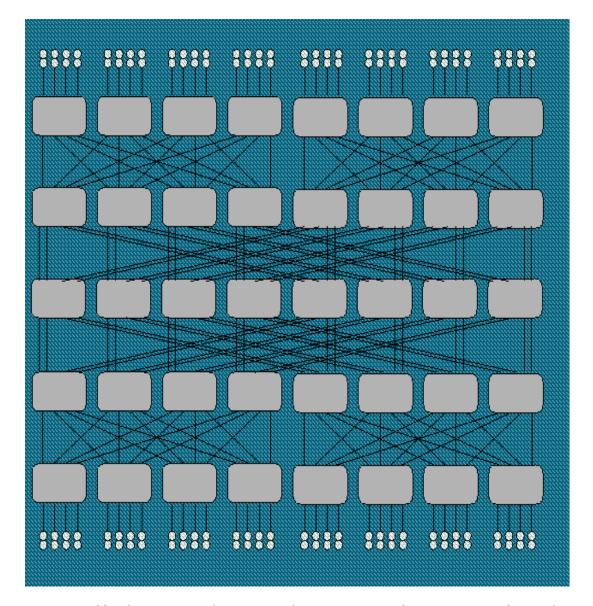


Research focuses on hierarchical systems:

-programming models-parallel libraries-performance tools-application porting-grid computing

http://zampano.zam.kfa-juelich.de

Bonn Parnass2 Cluster



144 Intel Pentium 2 400 MHz (dual), Myrinet

Applications:

- -parallel Navier-Stokes
- -molecular dynamics
- -algebraic multigrid
- -adaptive parallel multigrid
- -adaptive parallel sparse grids

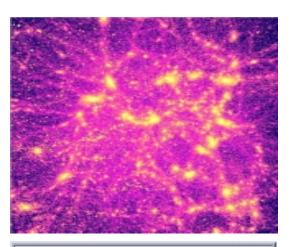
http://wissrech.iam.uni-bonn.de/research/projects/parnass2

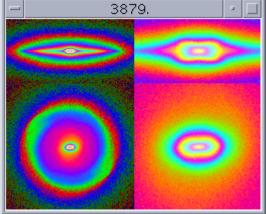
Geneva Gravitor II Cluster

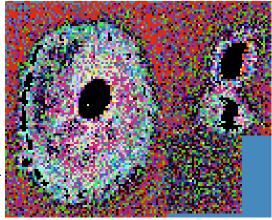
132 Intel Pentium 2 / Pentium 3 (heterogeneous)

Applications: Astrophysical Simulations

- -accretion of satellit galaxies
- -structure of the Milky Way
- -fusion of disk galaxies
- -pulsars search by data mining
- -formation of rotating stars
- -dark matter in galaxies



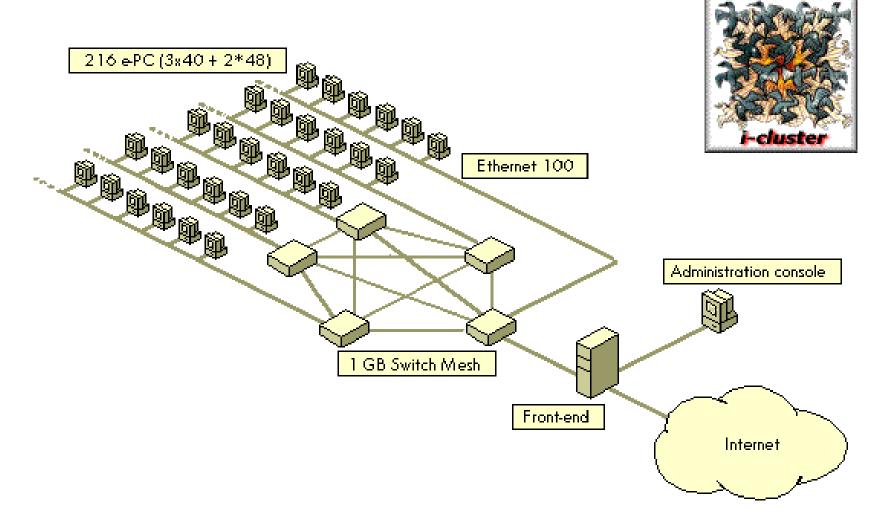




http://obswww.unige.ch/~pfennige/gravitor/gravitor_e.html

i-Cluster Grenoble

216 Intel Pentium 3 Nodes



http://www-id.imag.fr/Grappes/icluster/materiel.html

Conclusion

- Future developments in languages, compilers, and tools will support the transition to higher-level programming
- Hardware developments may include
 - new node architectures
 - massive parallelism with 1000s of nodes becoming standard
- A uniform and efficient high-level programming model for clusters with hierarchical parallelism is a research problem for some time to come