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Unstructured Scheduling in Parallel PDE Sparse Solvers on Distributed Memory Machines

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Report Number:
91-077

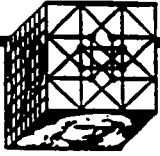
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**UNSTRUCTURED SCHEDULING IN PARALLEL PDE SPARSE
SOLVERS ON DISTRIBUTED MEMORY MACHINES**

**Mo Mu
John R. Rice**

**CSD-TR-91-077
November 1991**



**UNSTRUCTURED SCHEDULING
IN
PARALLEL PDE SPARSE SOLVERS
ON
DISTRIBUTED MEMORY MACHINES**

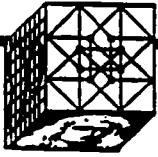
Mo Mu*
and
John R. Rice**

Computer Science Department
Purdue University
West Lafayette, IN 47907

October 25, 1991
Oak Ridge, TN

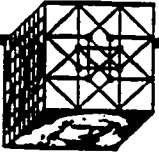
* Supported by NSF grant CCR-86-19817.

** Supported in part by AFOSR grant 88-0243 and
the Strategic Defense Initiative through ARO contract
DAAG03-90-0107.

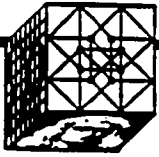


OUTLINE

- Background
- Underlying Algorithm
- Load Imbalance
- Unstructured Scheduling
- Other Optimization Strategies
- Conclusions

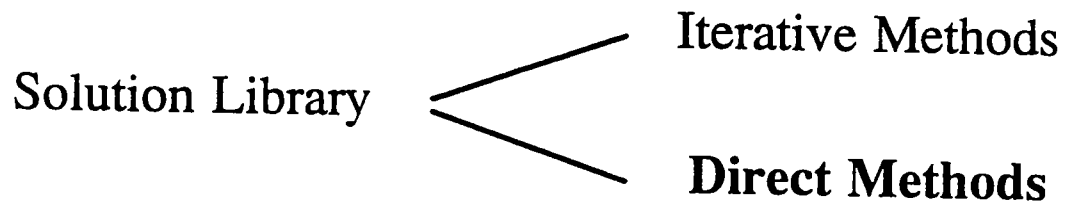


BACKGROUND

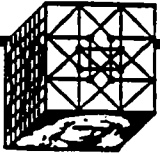


MOTIVATION

- Parallel ELLPACK

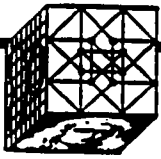


- Distributed memory machines



PDE PROBLEM

- General coefficients
- General boundary condition types
- General geometric domains



DISCRETIZATION

- Various Discretizations and Grids

Finite differences

Standard

High order

Finite elements

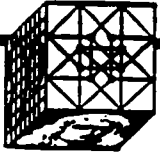
Collocation

Galerkin

on triangles or rectangles

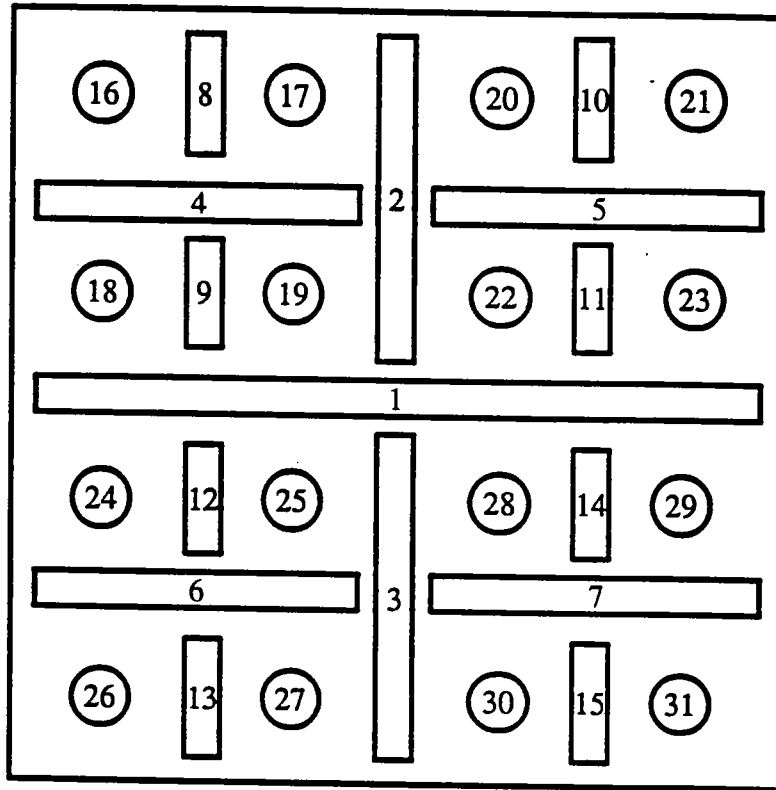
Hybrid schemes

- Distributed Over Processors



INDEXING

Incomplete Nested Dissection (domain decomposition based)

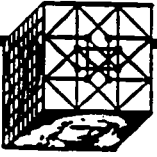


- *within each subdomain (“circle”)*

nested dissection
(potentially any efficient indexing scheme)

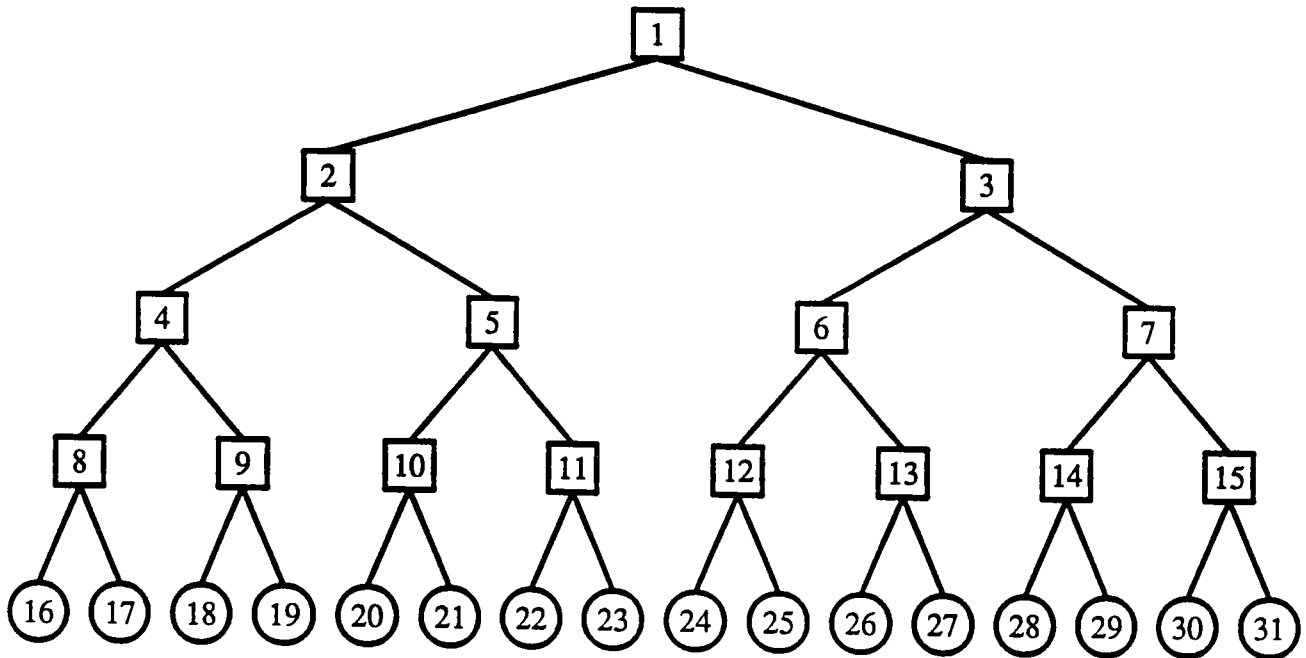
- *interface (the set of “boxes”)*

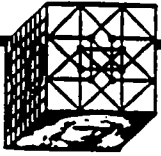
nested dissection



INDEXING (CONTINUED)

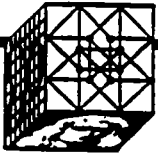
Elimination Tree





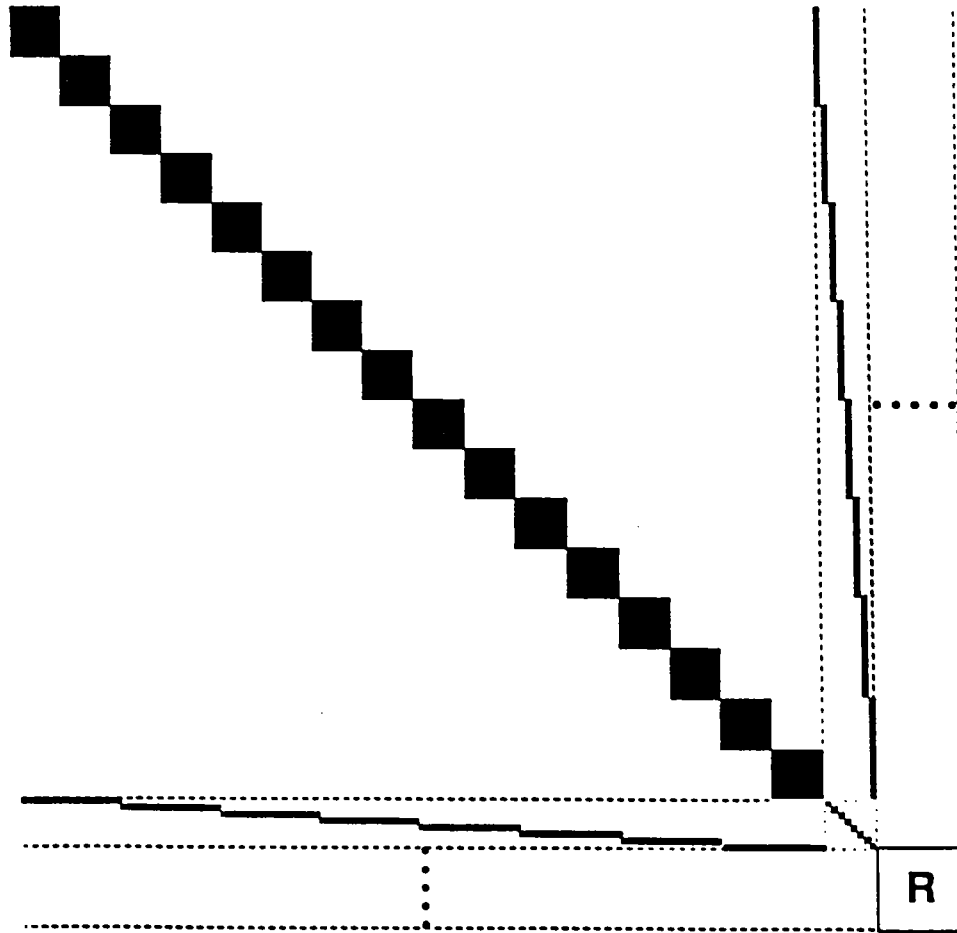
MATRIX PROBLEM

- Very large, sparse
- Nonsymmetric
- Block structured
- Distributed by row
- Numerically stable
- No symbolic factorization

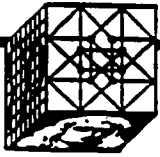


MATRIX PROBLEM (CONTINUED)

Sparse Matrix Structure



The sparse matrix structure for $p = 16$ processors. For the first two levels the solid boxes are where nonzero matrix elements might be (actually, these blocks are sparse also). The lower right box R contains diagonal blocks for the other 3 levels. Dots indicate sparse rows and columns. The relative sizes are correct for $n^2 = 100$, the number of grid points in one subdomain.



MATRIX PROBLEM (CONTINUED)

Sparse Matrix Structure

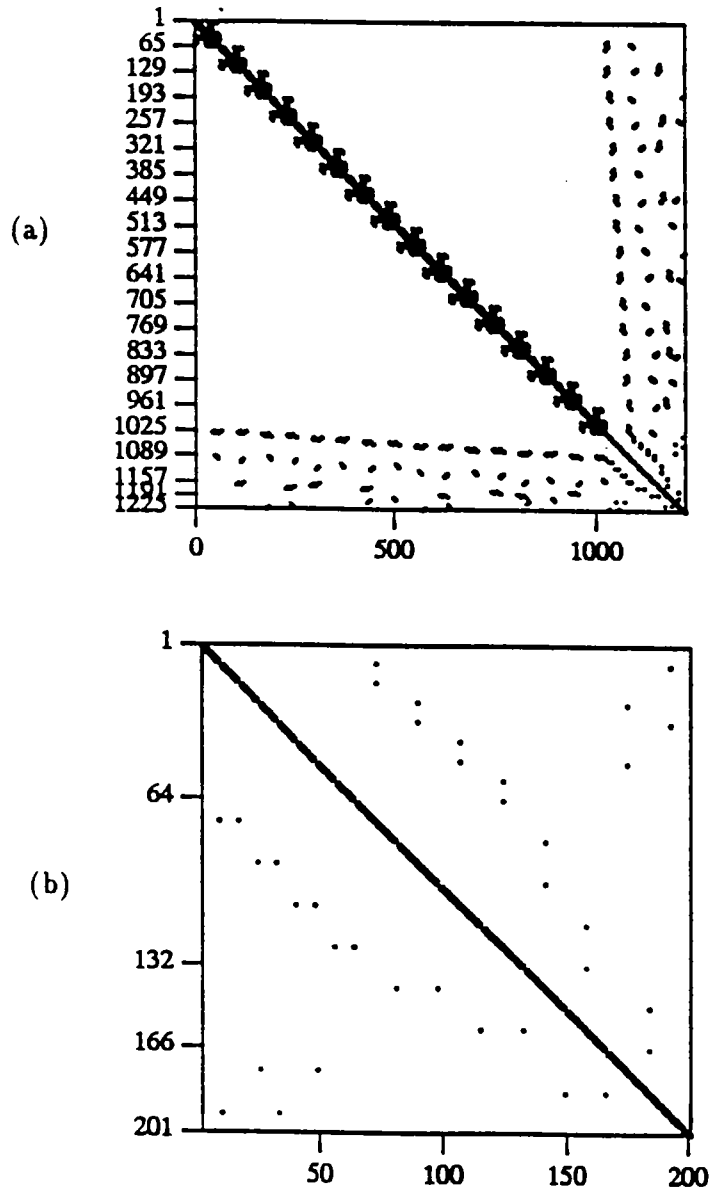
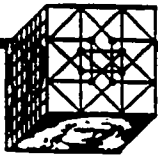
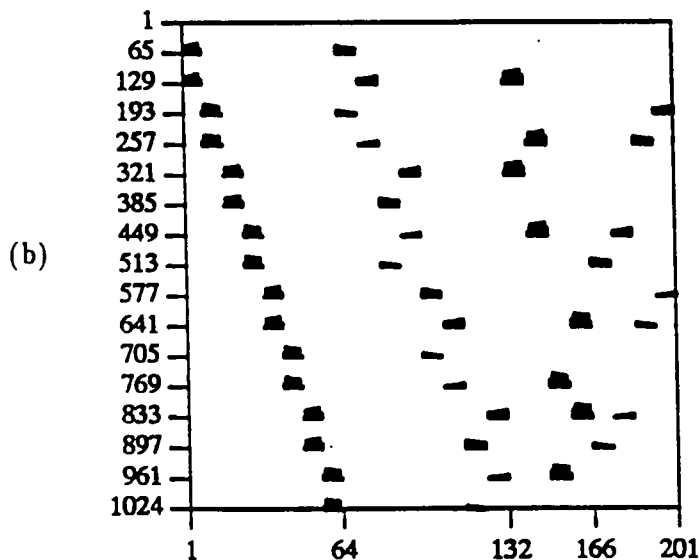
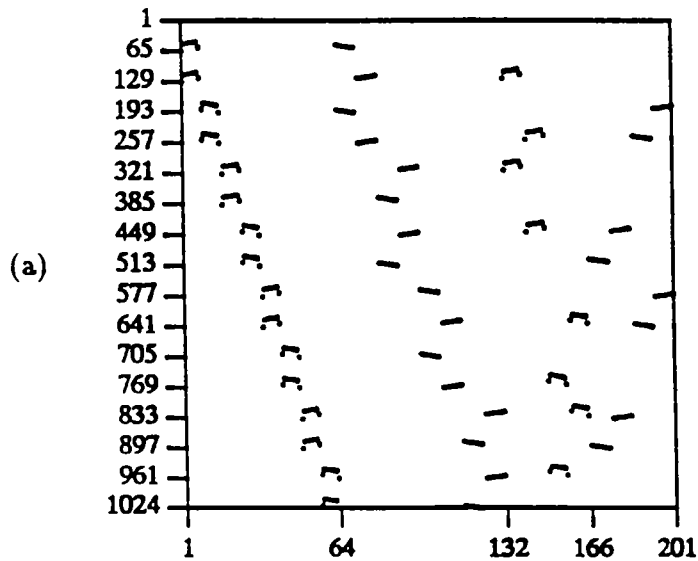


Figure 10: (a) Actual non-zero structure with $p = 16$, $n = 8$. The equation numbers are listed on the left. (b) The lower right block (everything except level 0) before the elimination starts.

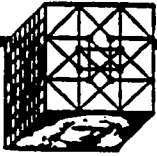


MATRIX PROBLEM (CONTINUED)

Sparse Matrix Structure



(a) The non-zero structure of the upper right matrix B before the elimination starts. Note that the display is distorted. B has 1024 rows and 201 columns. (b) The upper right matrix \bar{B} after the level 0 elimination.



MATRIX PROBLEM (CONTINUED)

Sparse Matrix Structure

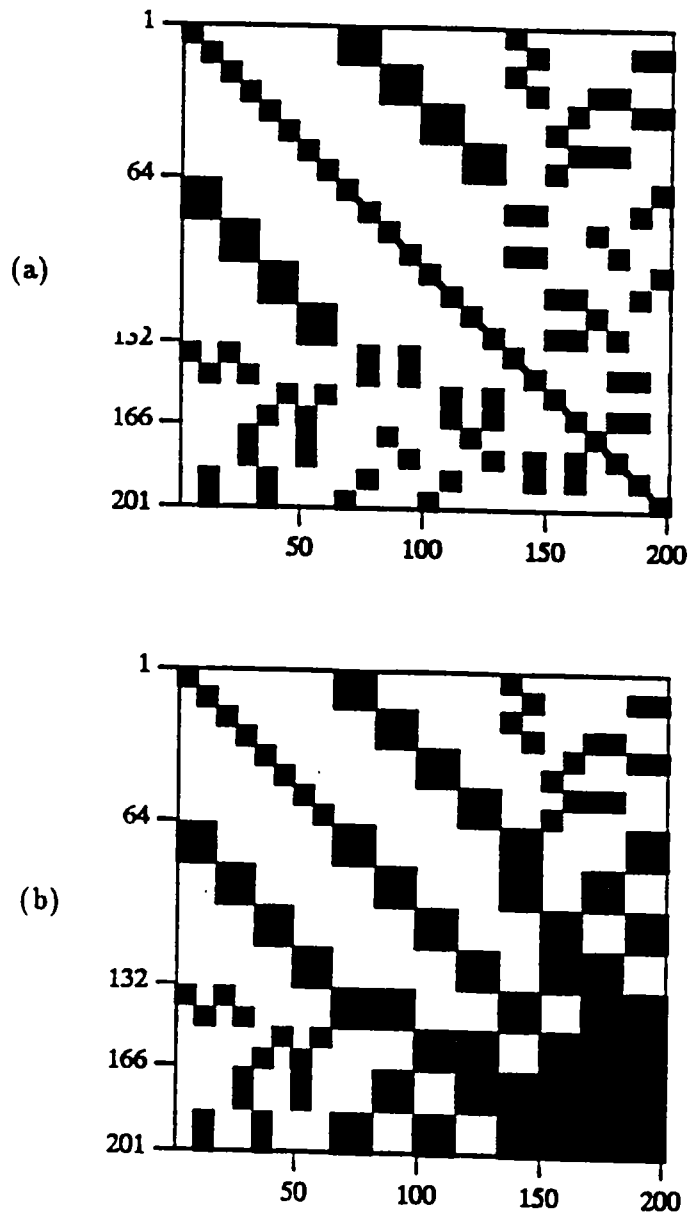
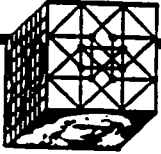
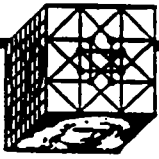


Figure 11: (a) The effect of the level 0 elimination on the lower right block. \bar{D} is given by (5). (b) The lower right block at the end of the elimination.

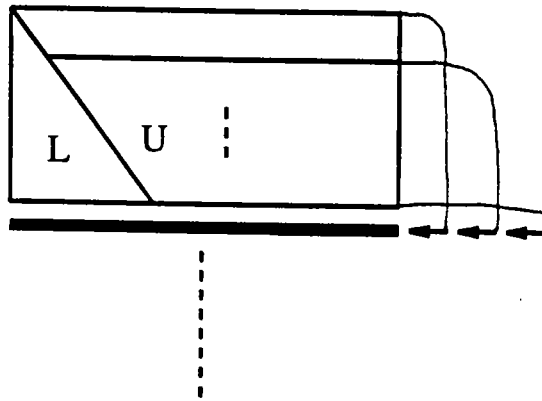


**UNDERLYING
ALGORITHM**



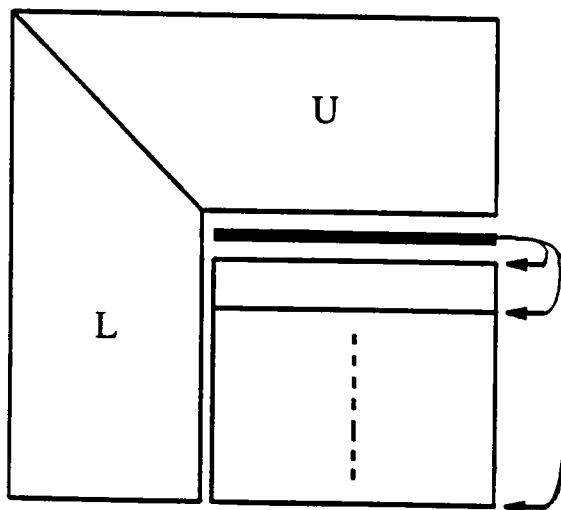
COMPUTATION ORGANIZATIONS

- up-looking

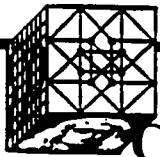


Do everything for an equation when you reach it.

- down-looking



Have the effects of elimination in an equation propagated before going on to the next equation.

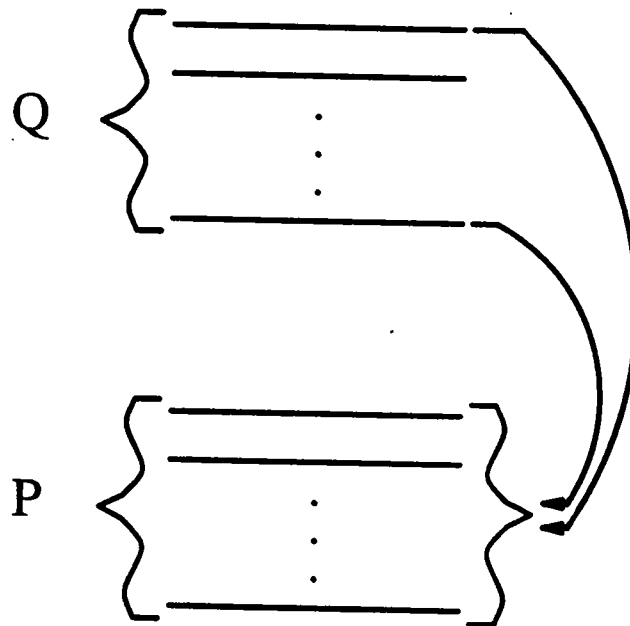


COMMUNICATION ORGANIZATIONS

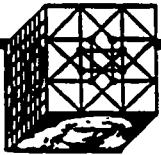
Q = Source

P = Destination

- fan-out



When processing an equation organize and pass on everything to later equations that they will need.

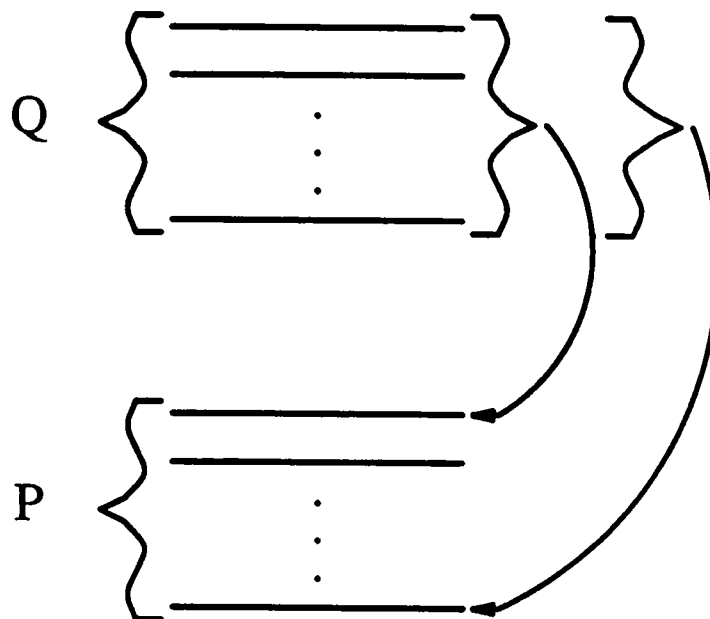


COMMUNICATION ORGANIZATIONS (CONTINUED)

Q = Source

P = Destination

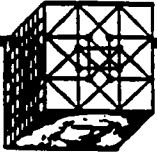
- fan-in



When processing an equation get everything from preceding equations that is needed.

$$\begin{aligned} Q : \mathbf{r}_i^q &= \sum_{k \in K} (a_{ik}/a_{kk}) * \text{row}_k \\ &= \sum_{k \in K} (a_{ki}/a_{kk}) * \text{row}_k \quad (\text{if } A \text{ is symmetric}) \end{aligned}$$

$$P : \text{row}_i = \text{row}_i - \mathbf{r}_i^q$$

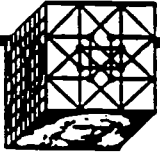


OBSERVATIONS AND FACTS

- Up-looking is better than down-looking in sparse data structure manipulation
- Fan-in has less communication overhead than fan-out
- Fan-out is suitable for down-looking
- Fan-in is suitable for up-looking
- Fan-in is not applicable to nonsymmetric matrices
 - (a) rows in the partial sum are in the source processor while the corresponding multipliers are in the destination processor;
 - (b) all multipliers of an equation in the destination processor have to be computed in a strictly sequential order by using rows distributed among various source processors

Possible way:

redistribute data and compute row i and column i at the same time



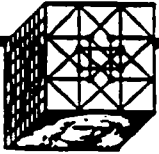
OUR SITUATION

Problem and Choice:

- Nonsymmetric matrices
- Fan-out communication organization
- Down-looking computation organization

Difficulties:

- Heavier communication overhead
- Communication buffer limit
- Destination list
- Up-looking used with fan-out requires a big storage buffer or repeated sending of same message.

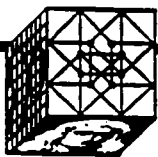


OUR APPROACH

Adapt ideas from other PDE solving methods, such as

- Domain Decomposition
- Substructuring

to direct sparse solvers



MATRIX FORMULATION

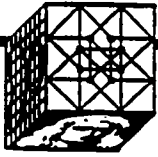
$$\begin{bmatrix} A_{11} & & & & & & B_1 \\ & A_{22} & & & & & B_2 \\ & & \cdot & & & & \cdot \\ & & & \cdot & & & \cdot \\ & & & & \cdot & & \cdot \\ & & & & & A_{pp} & B_p \\ C_1 & C_2 & C_1 & \dots & C_p & & D \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \cdot \\ \cdot \\ \cdot \\ x_p \\ x_d \end{bmatrix} = \begin{bmatrix} f_1 \\ f_2 \\ \cdot \\ \cdot \\ \cdot \\ f_p \\ f_d \end{bmatrix}$$

Schur Complement or Capacitance Matrix

$$S = D - \sum_{i=1}^p C_i A_{ii}^{-1} B_i$$

$$S x_d = f_d - \sum_{i=1}^p C_i A_{ii}^{-1} f_i$$

$$A_{ii} x_i = f_i - B_i x_d \quad i = 1, \dots, p$$

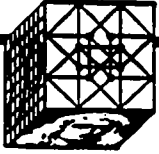


MAJOR STEPS

- factoring A_{ii}

$$A_{ii} = L_i U_i$$

- forming Schur Complement S
- factoring S



COMPUTING SCHUR COMPLEMENT

$$S = D - \sum_{i=1}^p C_i U_i^{-1} L_i^{-1} B_i$$

- Ordinary Gauss elimination algorithm

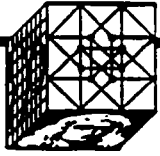
$$S = D - \sum_{i=1}^p (C_i U_i^{-1})(L_i^{-1} B_i)$$

- Implicit block factorization does not modify C_i matrices

$$S = D - \sum_{i=1}^p C_i (U_i^{-1} (L_i^{-1} B_i))$$

Advantages:

- sparsity of C_i matrices never lost
- reduced communication requirements similar to fan-in (next slide)
- static destination information is available from C_i matrices



COMPUTING SCHUR COMPLEMENT (CONTINUED)

Explicitly computing $A^{-1}B$ is too expensive!!!

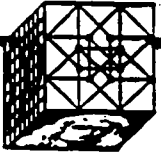
$$CA^{-1}B = \sum_k \text{col}_k(C) * \text{row}_k(A^{-1}B)$$

for ($\text{col}_k(C) \neq \text{null}$) **do**:

- solve $U^T y_k = e_k$ (triangular system of order $n - k + 1$)
- $\text{row}_k(A^{-1}B) = y_k^T(L^{-1}B)$

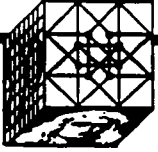
end k loop

- only subdomain boundary layer unknowns have $\text{col}_k(C) \neq \text{null}$, each of which corresponds to one communication with its partial sum (in the fan-in terminology, the modification vector, but it is much shorter here)
- very moderate increase in the computation overhead, which is compensated by the saving in the data structure manipulation for C
- flexible choices of ordering within the k -loop
- independent of local indexing



DATA STRUCTURES USED

- Subdomain equations — sparse
- Schur Complement — dense



ALGORITHMS

- **subdomains**

up-looking with “fan-in” type communication

- **interface**

down-looking with fan-out communication

Algorithm Outline

1. Apply up-looking Gauss elimination to sub-domain equations

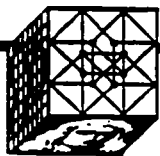
—— fully parallel

2. Participate in computing Schur Complement with “fan-in” type communication

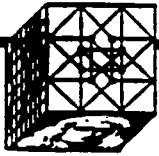
—— parallel and synchronized

3. Participate in factoring Schur Complement according to the elimination tree using down-looking with fan-out

—— parallel and synchronized



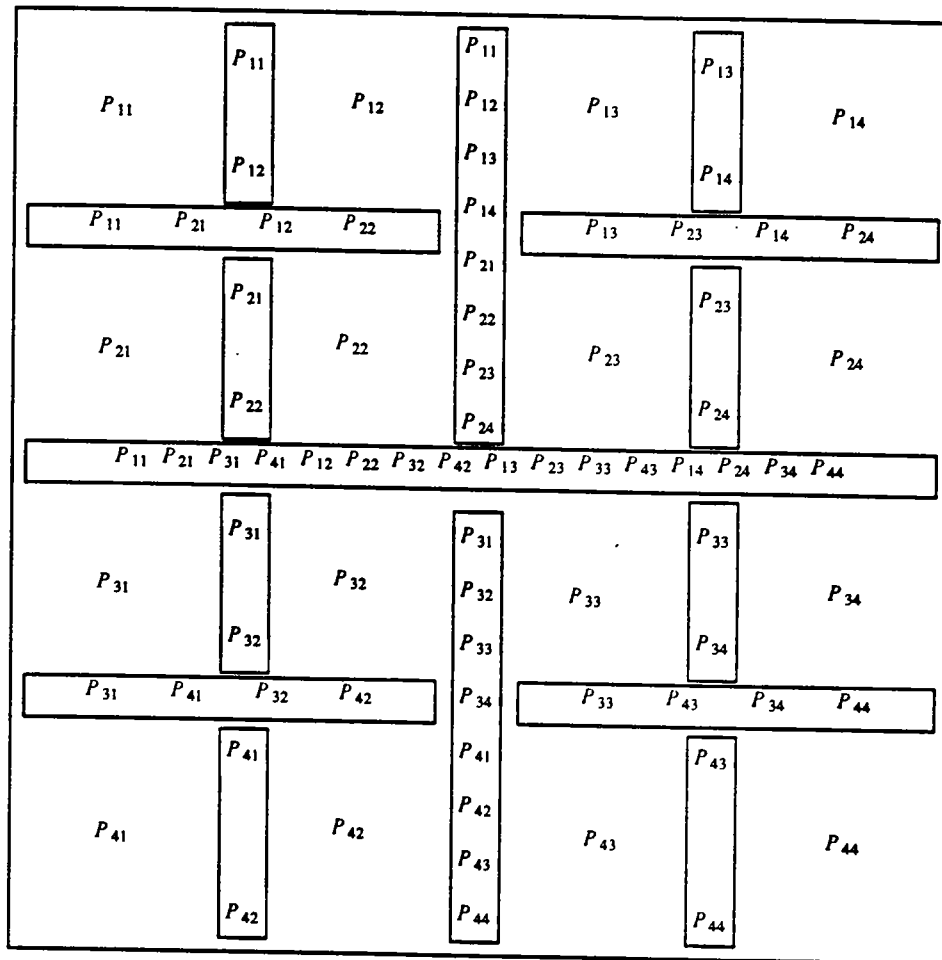
LOAD IMBALANCE



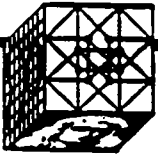
ASSIGNMENT

Equations to Processors

SUBCUBE-SUBTREE (Standard)

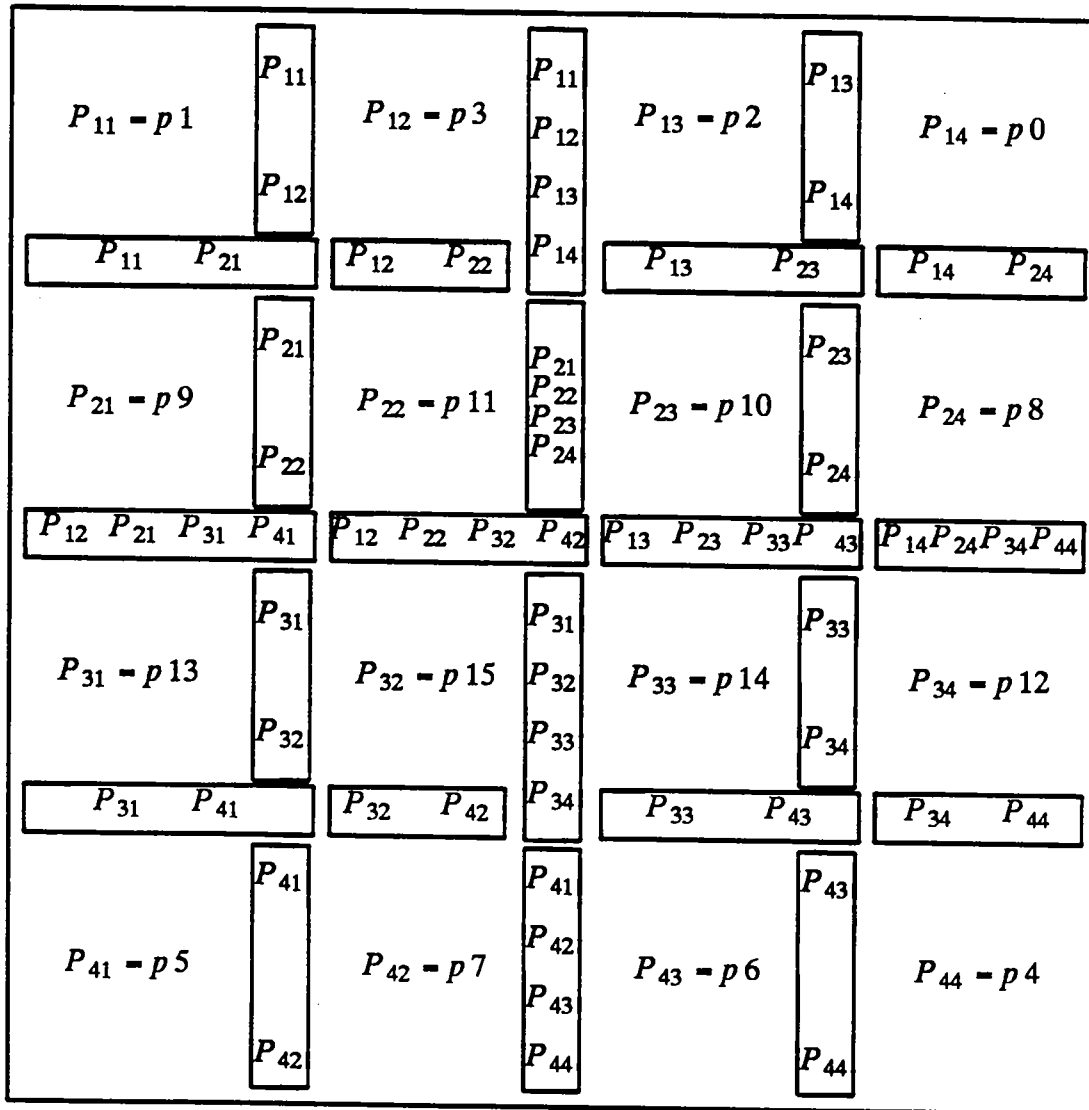


Standard subtree-subcube assignment for 16 processors. Within each box unknowns are assigned in wrapping manner to processors shown in the box.

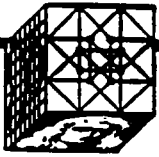


ASSIGNMENT (CONTINUED)

GRID-SUBCUBE-SUBTREE (Grid)



Grid based subtree-subcube assignment for 16 processors. Within the subdomain interfaces we show how the processors are assigned to unknowns in parts of the separators.



PERFORMANCE, 16 PROCESSORS

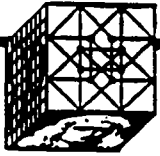
- on the NCUBE/2

| Grid | Sequential time | Parallel time | Speedup |
|---------|-----------------|---------------|---------|
| 21 × 21 | 0.578 | 0.118 | 4.90 |
| 25 × 25 | 1.05 | 0.173 | 6.07 |
| 29 × 29 | 1.77 | 0.244 | 7.25 |
| 33 × 33 | 2.73 | 0.340 | 8.03 |
| 37 × 37 | 4.03 | 0.489 | 8.24 |
| 41 × 41 | 5.69 | 0.659 | 8.63 |
| 45 × 45 | 7.73 | 0.843 | 9.17 |
| 49 × 49 | 10.23 | 1.07 | 9.56 |
| 53 × 53 | 13.21 | 1.397 | 9.46 |
| 57 × 57 | 16.78 | 1.75 | 9.59 |
| 61 × 61 | 20.87 | 2.09 | 9.98 |
| 65 × 65 | 25.67 | 2.46 | 10.43 |

- on the Intel i860

| Grid | Sequential time | Parallel time | Speedup |
|---------|-----------------|---------------|-----------------|
| 21 × 21 | 0.071 | 0.094 | xxx |
| 57 × 57 | 1.87 | 0.673 | 2.78 |

2.91



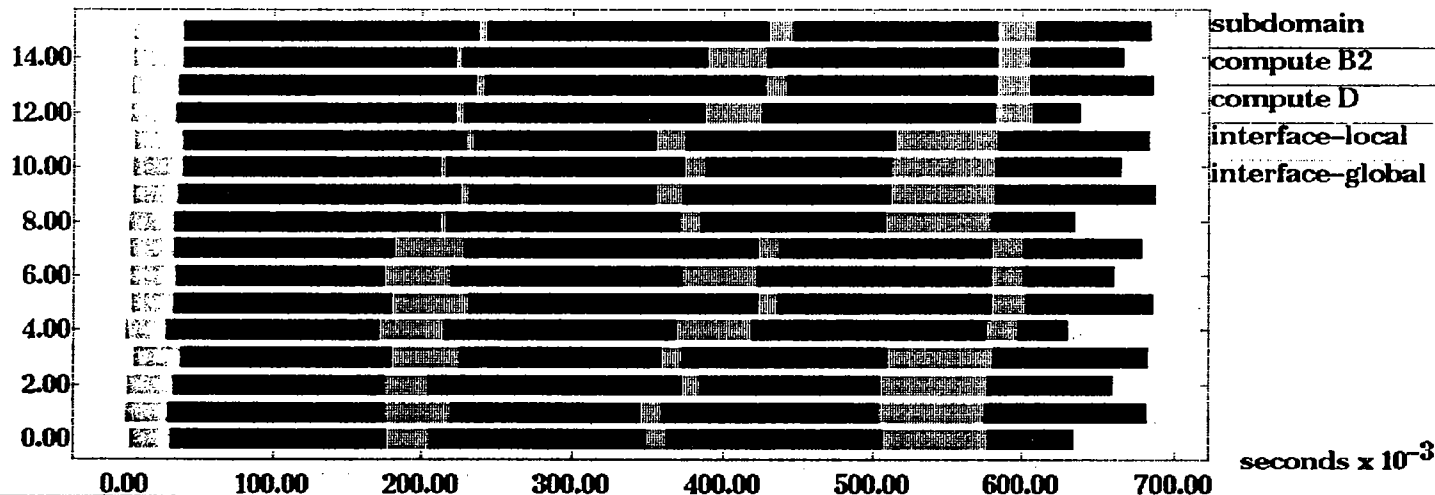
VISUALIZING PERFORMANCE

- subdomain — almost load balanced
- $A^{-1}B$ — very unbalanced
- $CA^{-1}B$ — a lot of idle time
- interface — a lot of synchronization
- sending message — substantial overhead
on the Intel i860
- varying grid — similar performance behavior

Close Hardcopy About

EXECUTION PHASES ON Intel i860

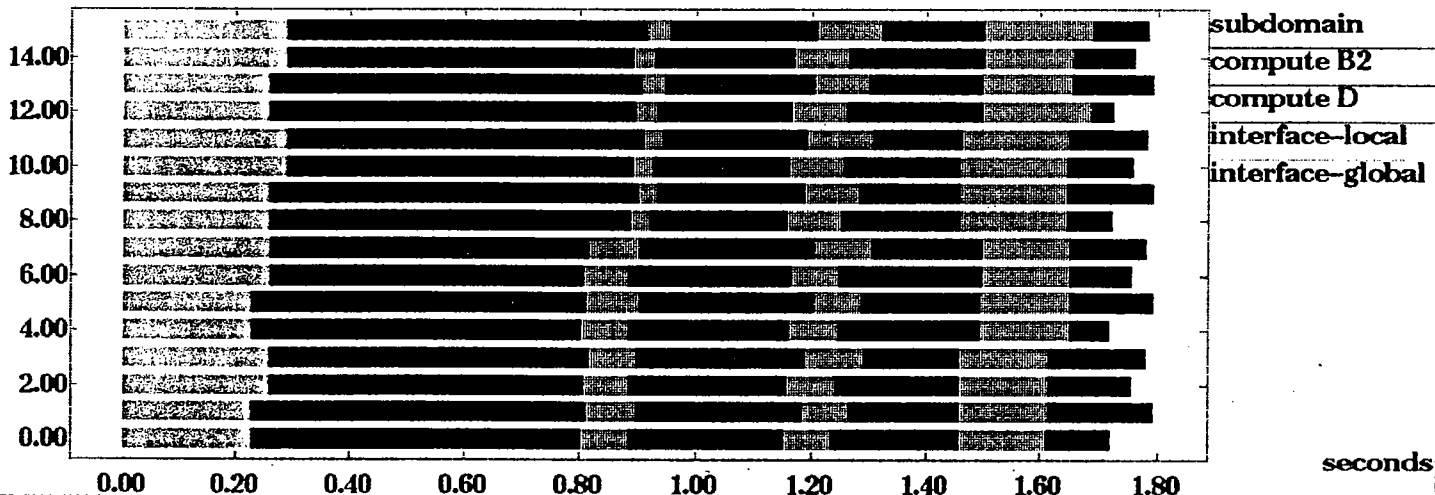
processor



Close Hardcopy About

EXECUTION PHASES ON NCUBE/2

processor

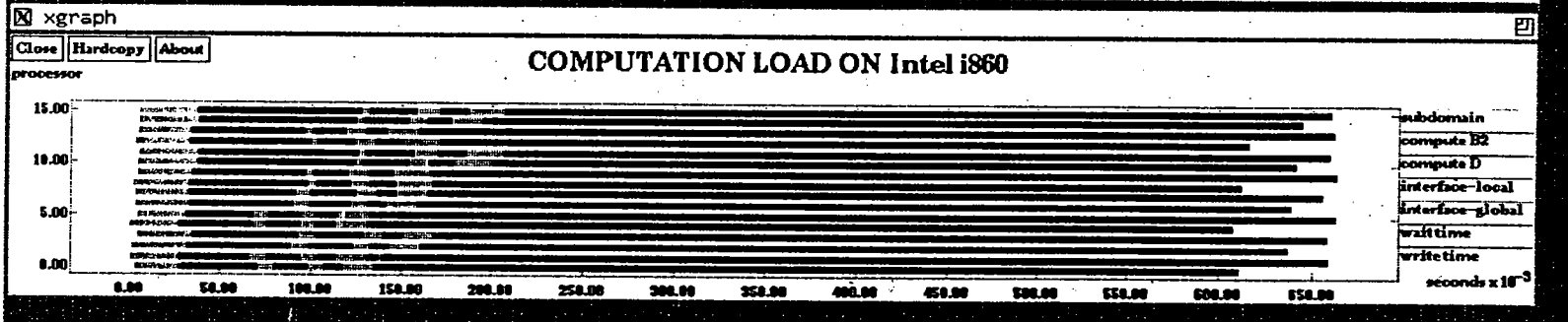
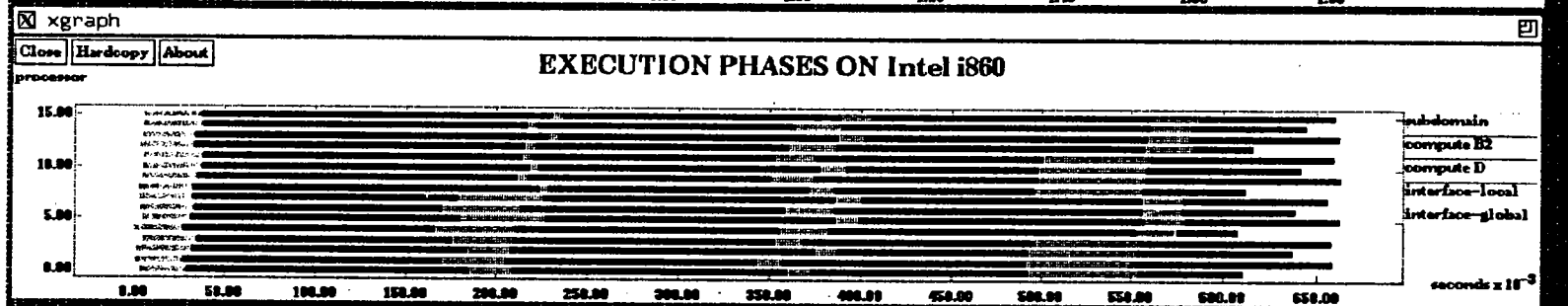
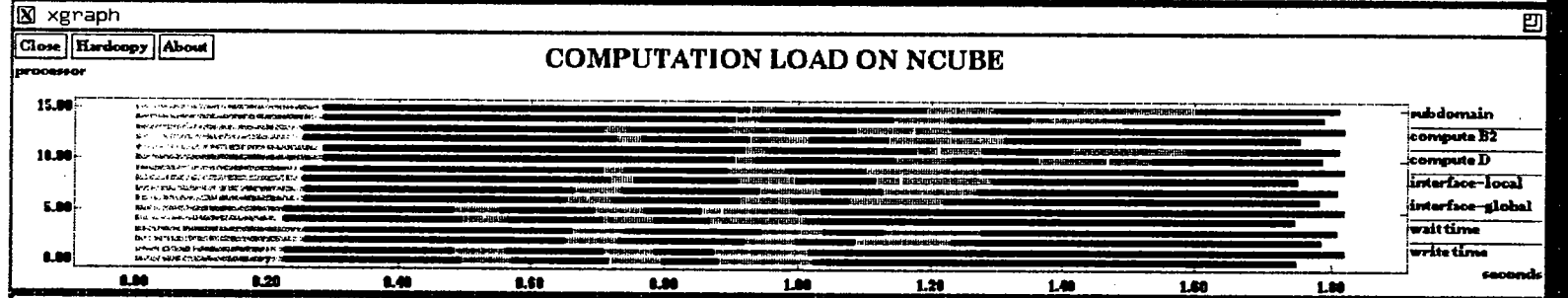
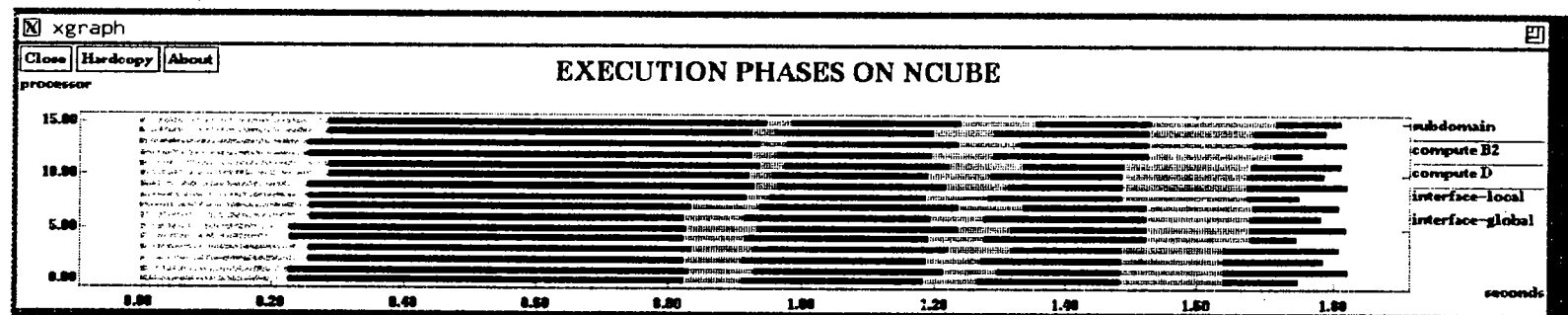


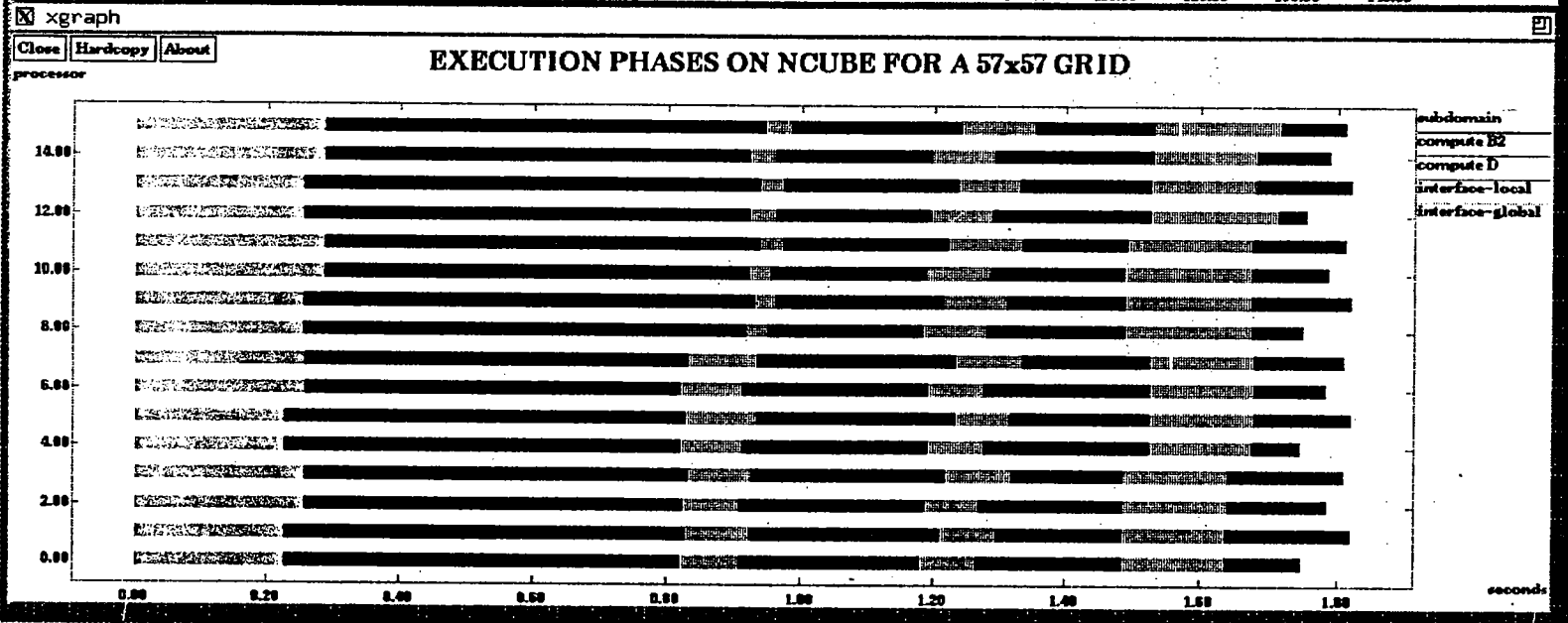
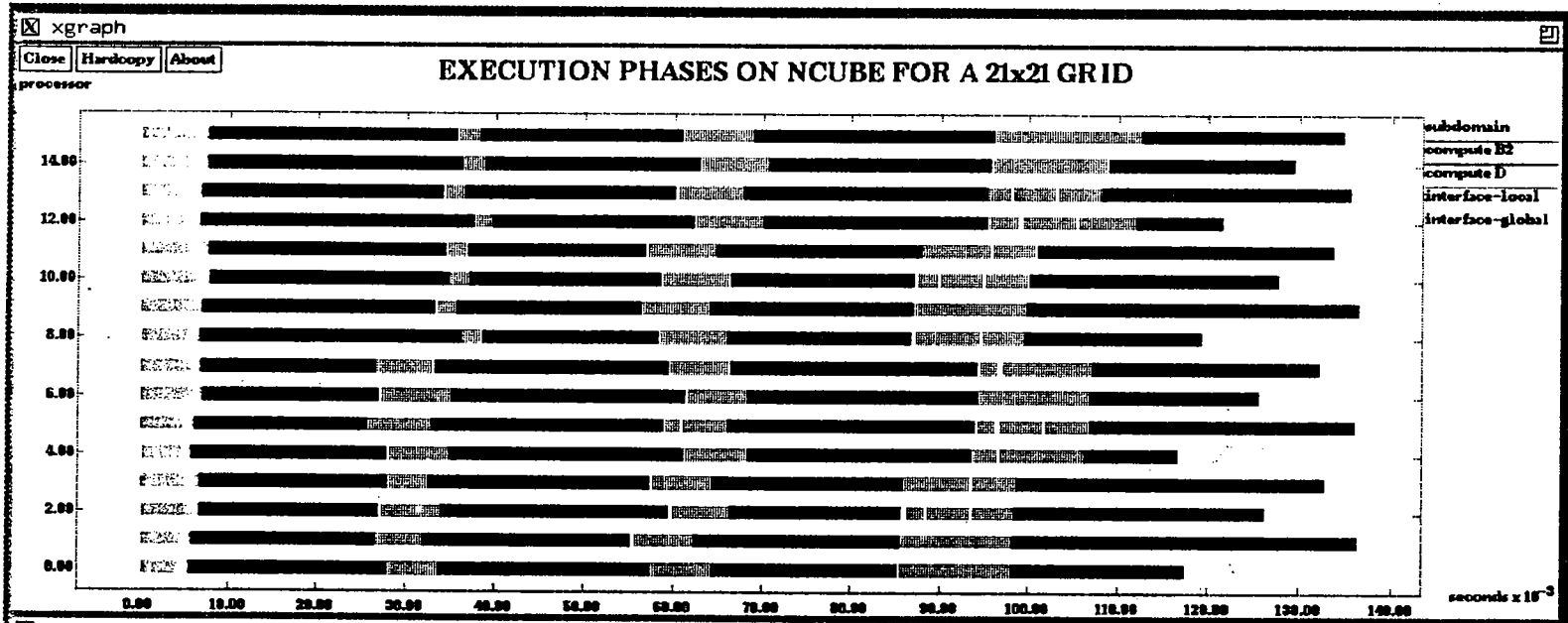
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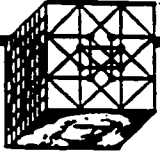


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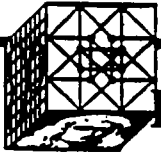
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UNSTRUCTURED SCHEDULING



REORGANIZE COMPUTATION AND COMMUNICATION IN FORMING SCHUR COMPLEMENT

To reduce synchronization time, compute rows of $A^{-1}B$ in an order that sends work first to idle processors using the following priorities.

- priority 1 — corner processors:

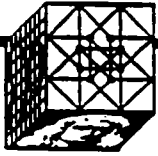
P0, P1, P4 and P5

- priority 2 — other border processors:

P2, P3, P6, P7, P8, P9, P12, P13

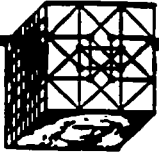
- priority 3 — center processors:

P10, P11, P14, P15

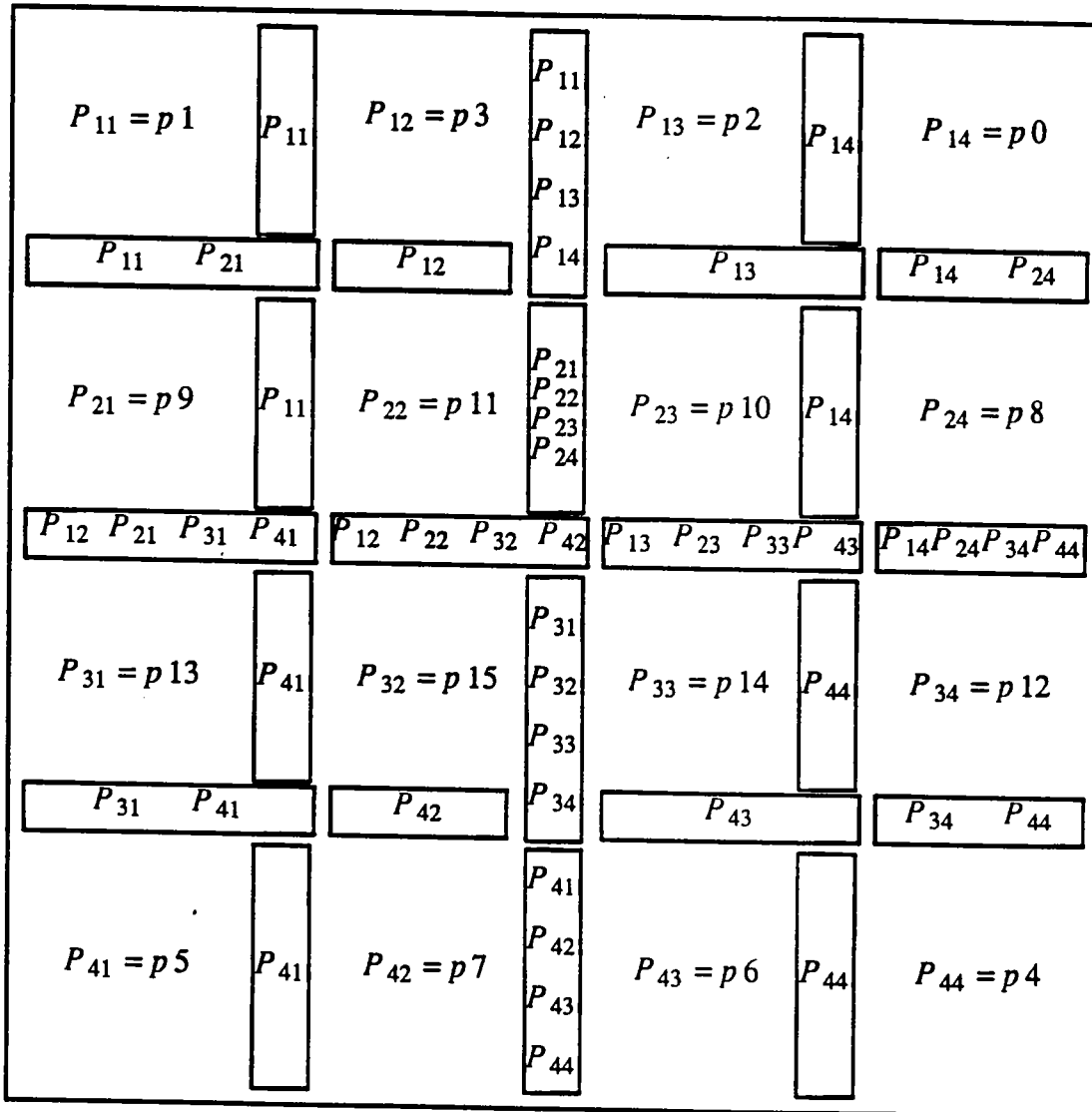


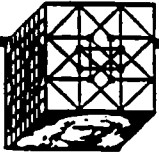
REASSIGN THE DATA AND TASKS

- move tasks from busy processors to idle processors
- overlap computation and communication



REASSIGNMENT





EFFECTS OF RESCHEDULING

- On the NCUBE/2

57 × 57 grid:

| | |
|---------------|--------------|
| parallel time | 1.75 → 1.54 |
| speedup | 9.59 → 10.89 |

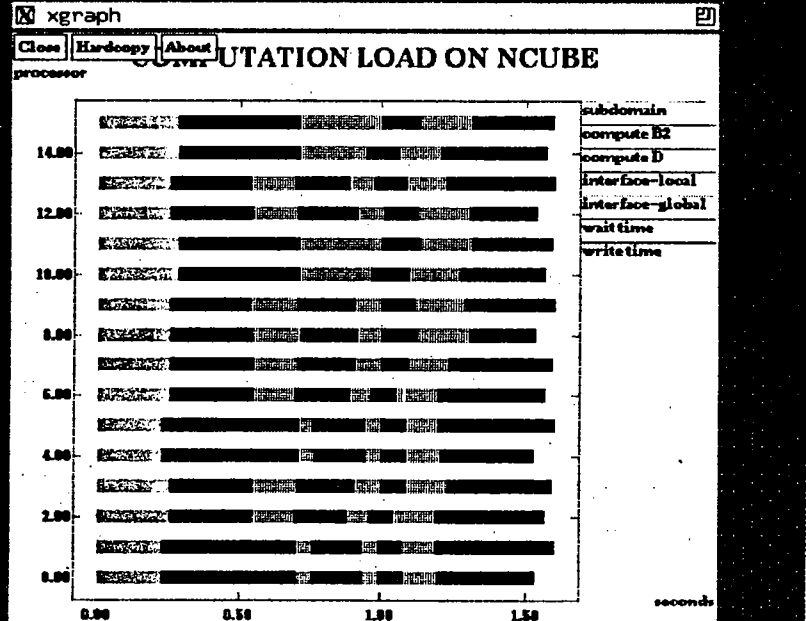
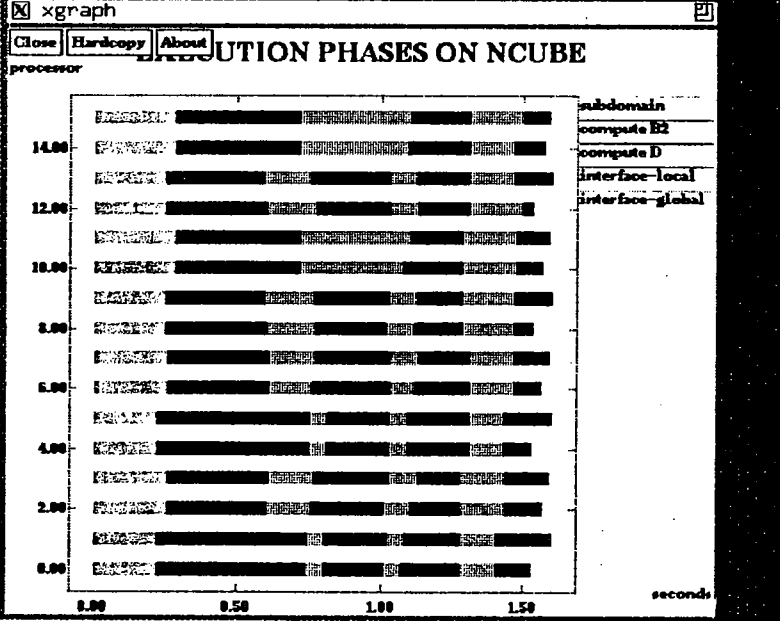
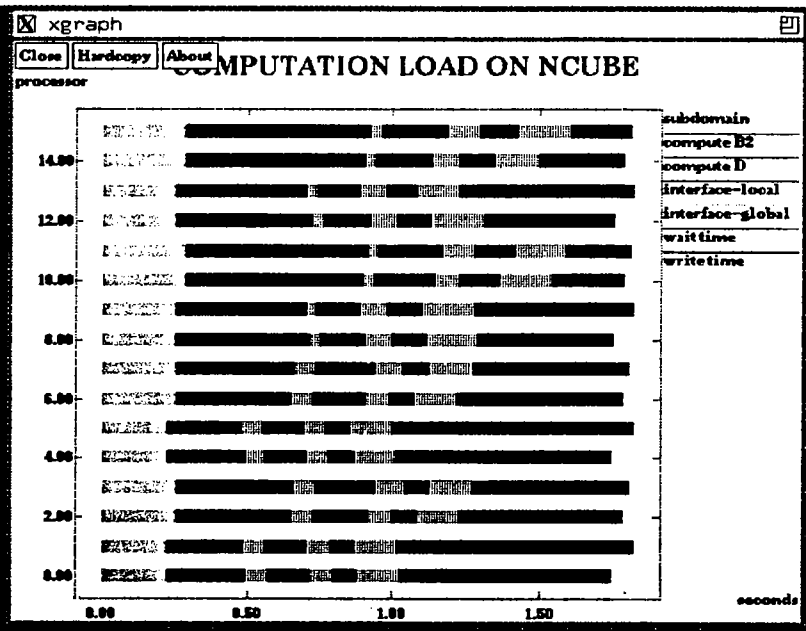
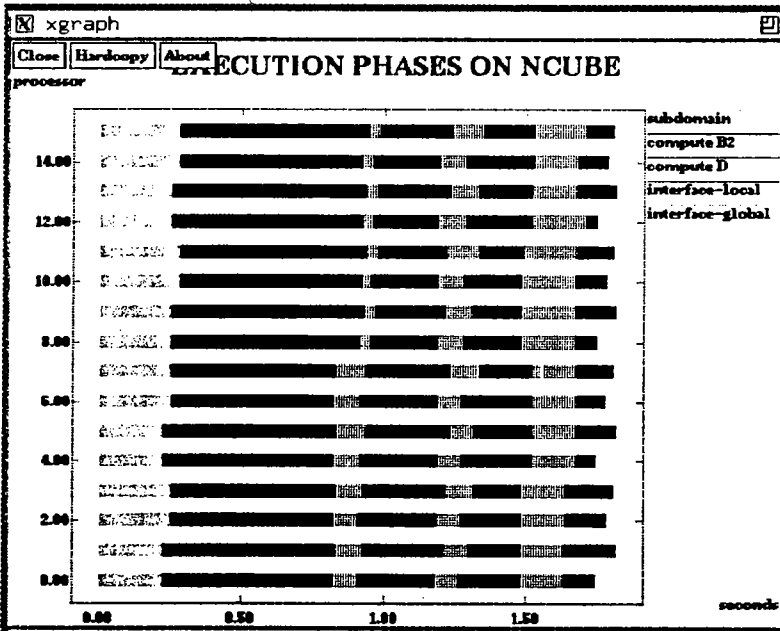
61 × 61 grid:

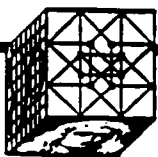
| | |
|---------------|--------------|
| parallel time | 2.09 → 1.87 |
| speedup | 9.98 → 11.15 |

- On the i860

no improvement

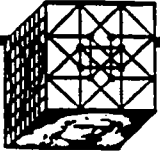
- (a) the effect of communication dominates that of the load imbalance too much
- (b) heavy overhead of sending message



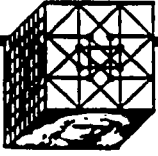


OPTIMAL SCHEDULINGS

- Very unstructured
- Mutual interactions of load balancing in rescheduling and synchronization in computing S
- Coarse grid analysis



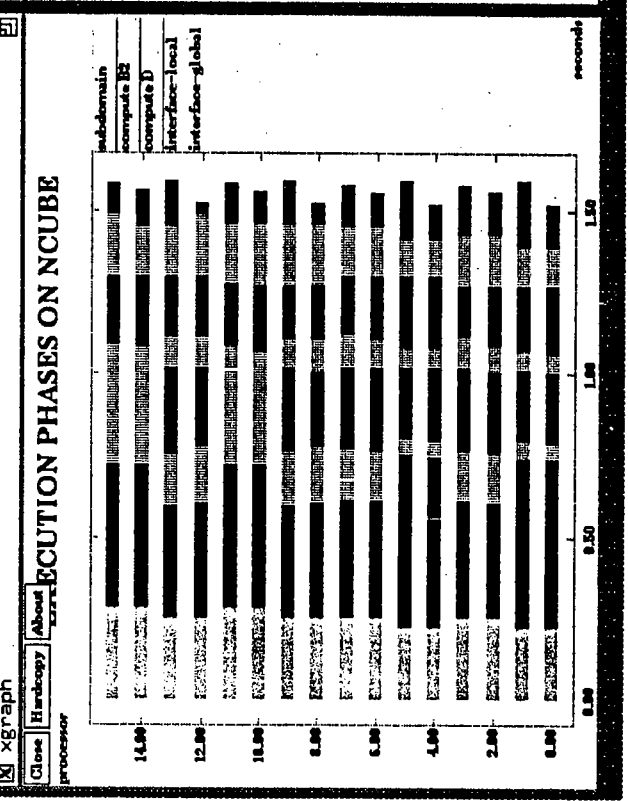
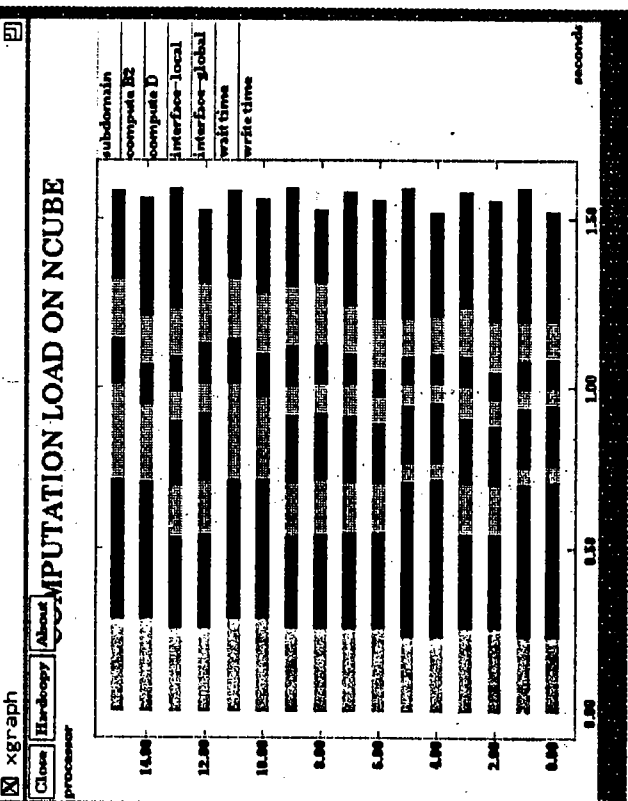
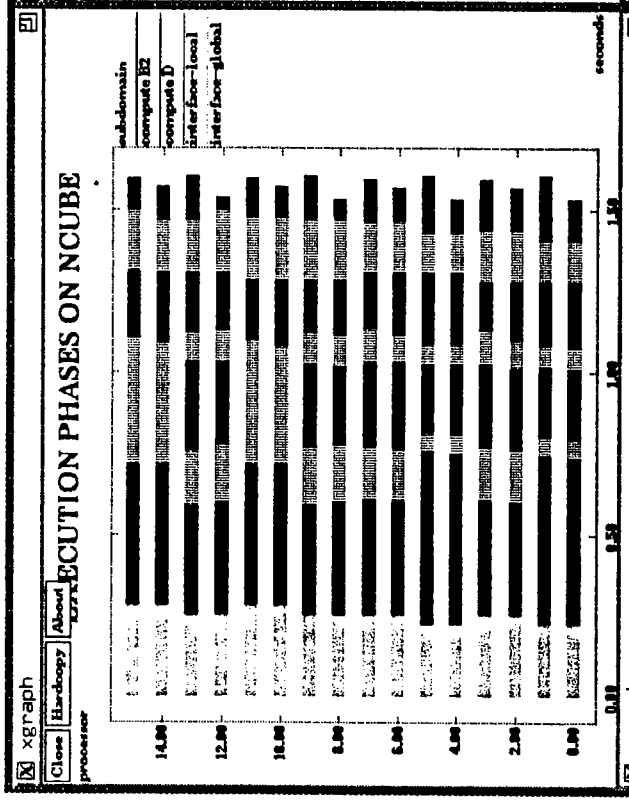
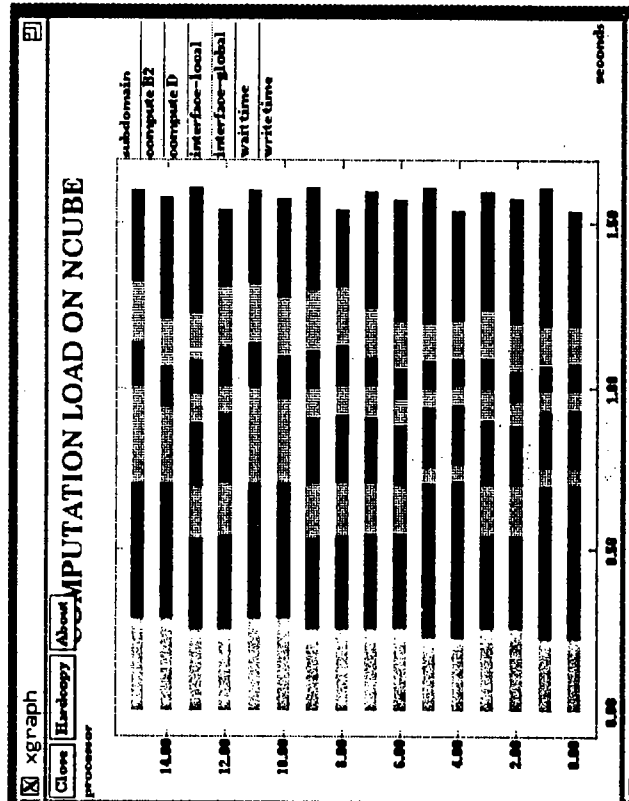
OTHER OPTIMIZATION STRATEGIES

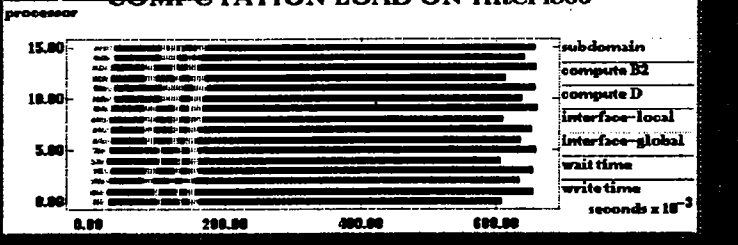
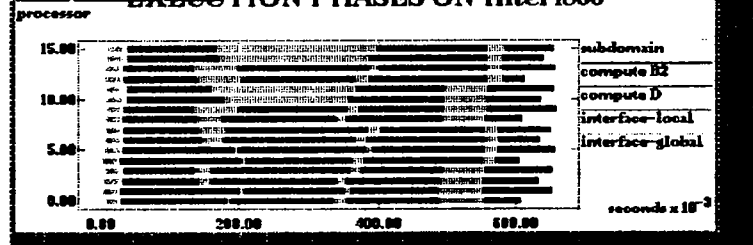
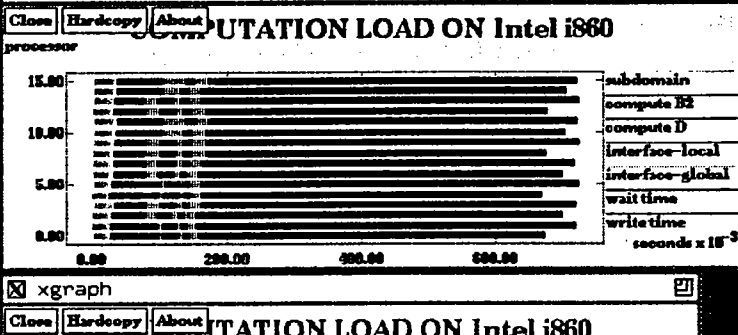
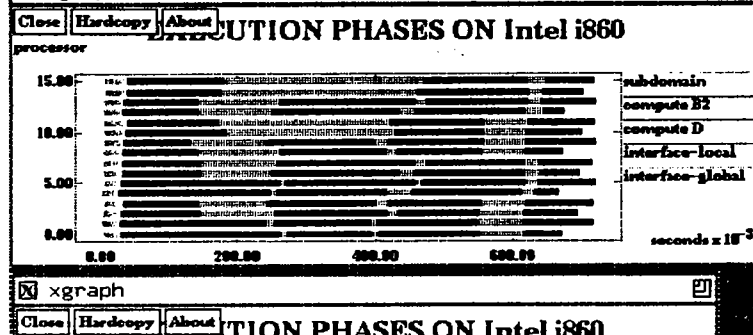
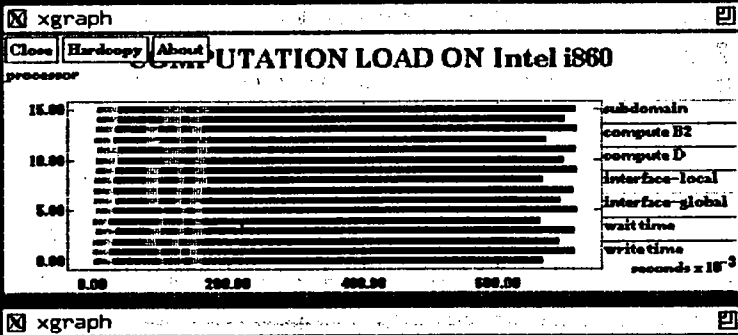
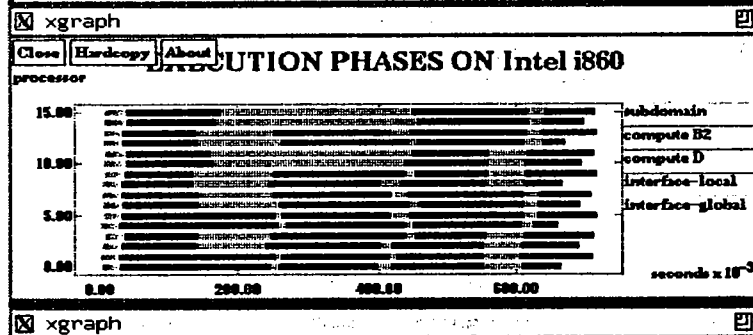
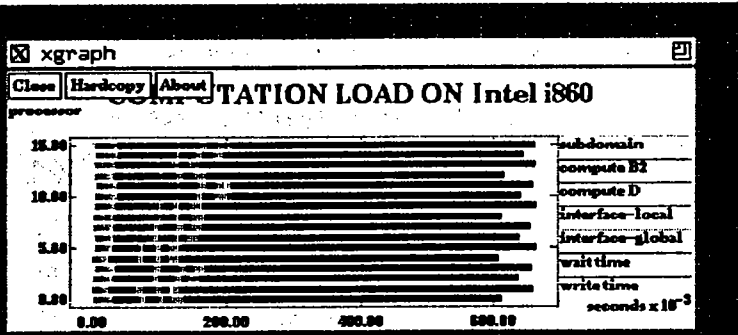
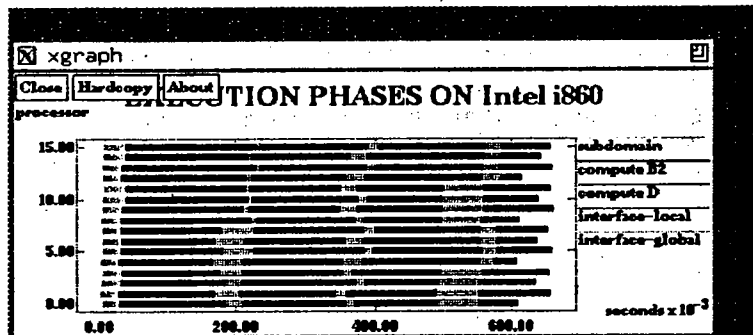


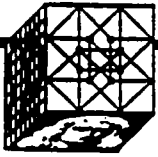
PACKING VS. PIPELINING

- Pack messages when pipelining is not important
- Trade-off between packing and pipelining by adjusting a `grain_control` parameter in rescheduling

Slide 35a

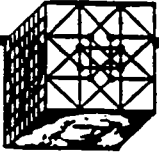






OTHER STRATEGIES

- Replace multicast by broadcast when the remaining matrix becomes much denser
- Use irregular grids



CONCLUSIONS

- The parallel PDE sparse solver is load unbalanced with the standard scheduling
- The parallel PDE sparse solver can gain high speedup by reorganizing and overlapping computation and communication using proper schedulings
- The i860 machine is an unbalanced design for many more scientific applications than the NCUBE 2 or Intel iPSC/2