## High Performance Numerical Libraries

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## Gaussian Elimination - Review

## Version 1

for each column i
zero it out below the diagonal by adding multiples of row $i$ to later rows
for $\mathrm{i}=1$ to $\mathrm{n}-1$
for each row j below row i
for $\mathrm{j}=\mathrm{i}+1$ to n
add a multiple of row $i$ to row $j$
for $k=i$ to $n$

$$
A(j, k)=A(j, k)-A(j, i) / A(i, i) * A(i, k)
$$



## Gaussian Elimination - Review

## Version 2 - Remove $A(j, i) / A(i, i)$ from inner loop

 for each column izero it out below the diagonal by adding multiples of row $i$ to later rows

## for $\mathrm{i}=1$ to $\mathrm{n}-1$

for each row j below row i
for $\mathrm{j}=\mathrm{i}+1$ to n
$m=A(j, i) / A(i, i)$
for $k=i$ to $n$

$$
A(j, k)=A(j, k)-m^{*} A(i, k)
$$



## Gaussian Elimination - Review

## Version 3 - Don't compute what we already know

 for each column izero it out below the diagonal by adding multiples of row $i$ to later rows

## for $\mathrm{i}=1$ to $\mathrm{n}-1$

for each row j below row i
for $\mathrm{j}=\mathrm{i}+1$ to n
$m=A(j, i) / A(i, i)$
for $k=i+1$ to $n$

$$
A(j, k)=A(j, k)-m^{*} A(i, k)
$$



## Gaussian Elimination - Review

## Version 4 - Store multipliers m below diagonals

 for each column izero it out below the diagonal by adding multiples of row $i$ to later rows

## for $\mathrm{i}=1$ to $\mathrm{n}-1$

for each row j below row i
for $j=i+1$ to $n$
$A(j, i)=A(j, i) / A(i, i)$
for $k=i+1$ to $n$
$A(j, k)=A(j, k)-A(j, i)^{*} A(i, k)$


## GE - Runtime

$\square$ Divisions

$$
1+2+3+\ldots(n-1)=n^{2} / 2 \text { (approx.) }
$$

$\square$ Multiplications / subtractions

$$
1^{2}+2^{2}+3^{2}+4^{2}+5^{2}+\ldots .(n-1)^{2}=n^{3 / 3}-n^{2} / 2
$$

$\square$ Total

$$
2 n^{3} / 3
$$

## Parallel GE

$\square 1^{\text {st }}$ step -1 -D block partitioning along blocks of $n$ columns by $p$ processors


## 1D block partitioning - Steps

## 1. Divisions

$$
n^{2} / 2
$$

2. Broadcast

$$
\begin{aligned}
& x \log (p)+y \log (p-1)+z \log (p-3)+\ldots \log 1< \\
& n^{2} \log p
\end{aligned}
$$

3. Multiplications and Subtractions

$$
(n-1) n / p+(n-2) n / p+\ldots .1 \times 1=n^{3} / p \text { (approx.) }
$$

Runtime:

$$
<n^{2} / 2+n^{2} \log p+n^{3} / p
$$

## 2-D block

## $\square$ To speedup the divisions



## 2D block partitioning - Steps

## 1. Broadcast of $(k, k)$

 $\log Q$2. Divisions
n²/Q (approx.)
3. Broadcast of multipliers

$$
x \log (P)+y \log (P-1)+z \log (P-2)+\ldots .=n^{2} / Q \log P
$$

4. Multiplications and subtractions
$n^{3} / P Q$ (approx.)

Problem with block partitioning for GE

- Once a block is finished, the corresponding processor remains idle for the rest of the execution
$\square$ Solution? -


## Onto cyclic

$\square$ The block partitioning algorithms waste processor cycles. No load balancing throughout the algorithm.
$\square$ Onto cyclic

cyclic
Load balance


1-D block-cyclic
Load balance, block operations, but column factorization bottleneck

| 0 | 1 | 0 | 1 | 0 | 1 | 0 | 1 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| 2 | 3 | 2 | 3 | 2 | 3 | 2 | 3 |
| 0 | 1 | 0 | 1 | 0 | 1 | 0 | 1 |
| 2 | 3 | 2 | 3 | 2 | 3 | 2 | 3 |
| 0 | 1 | 0 | 1 | 0 | 1 | 0 | 1 |
| 2 | 3 | 2 | 3 | 2 | 3 | 2 | 3 |
| 0 | 1 | 0 | 1 | 0 | 1 | 0 | 1 |
| 2 | 3 | 2 | 3 | 2 | 3 | 2 | 3 |

2-D block-cyclic
Has everything

## Block cyclic

$\square$ Having blocks in a processor can lead to block-based operations (block matrix multiply etc.)
$\square$ Block based operations lead to high performance

## GE: Miscellaneous GE with Partial Pivoting

$\square$ 1D block-column partitioning: which is better? Column or row pivoting
-Column pivoting does not involve any extra steps since pivot search and exchange are done locally on each processor. $\mathrm{O}(\mathrm{n}-\mathrm{i}-1)$
-The exchange information is passed to the other processes by piggybacking with the multiplier information

- Row pivoting
- Involves distributed search and exchange $-O(n / P)+O(\log P)$
$\square$ 2D block partitioning: Can restrict the pivot search to limited number of columns


## Sparse Iterative Methods

# Iterative \& Direct methods - Pros and Cons. 

$\square$ Iterative methods do not give accurate results.
$\square$ Convergence cannot be predicted
$\square$ But absolutely no fills.

## Parallel Jacobi, Gauss-Seidel, SOR

$\square$ For problems with grid structure (1D, 2-D etc.), Jacobi is easily parallelizable
$\square$ Gauss-Seidel and SOR need recent values. Hence ordering of updates and sequencing among processors
$\square$ But Gauss-Seidel and SOR can be parallelized using red-black ordering or checker board

## 2D Grid example



## Red-Black Ordering

$\square$ Color alternate nodes in each dimension red and black
$\square$ Number red nodes first and then black nodes
$\square$ Red nodes can be updated simultaneously followed by simultaneous black nodes updates

## 2D Grid example - Red Black Ordering


$\square$ In general, reordering can affect convergence


## Graph Coloring

$\square$ In general multi-colored graph coloring Ordering for parallel computing of GaussSeidel and SOR

