# Resource-aware Distributed Block-based LU Decomposition on Wireless Sensor Networks



## Introduction

In this work, we describe the implementation of a distributed LU decomposition on wireless sensor networks. It forms part of a numerical methods *kernel* for distributed signal processing on wireless sensor networks.

#### Contribution

> Algorithm for distributing the LU decomposition of an NxN matrix

- Energy and storage-aware task allocation
- > Fast and smooth recovery from node failure
- Efficient scheduling and time slot allocation
- > Exhaustive analysis and experiments on Telosb platform

#### Background

LU decomposition of a 3x3 matrix involves finding lower and upper triangular matrices such that L.U = A:

 $\begin{bmatrix} a_{11} & 0 & 0 \\ a_{21} & a_{22} & 0 \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \cdot \begin{bmatrix} \beta_{11} & \beta_{12} & \beta_{13} \\ 0 & \beta_{22} & \beta_{23} \\ 0 & 0 & \beta_{33} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$ 

Crout's algorithm for finding the LU decomposition of a 3x3:

1) Set 
$$a_{ii} = 1 \quad \forall i = 1,...,N$$
  
2) solve for  $\beta_{ij}$  and  $\alpha_i \quad \beta_{ij} = a_{ij} - \sum_{\substack{k=1\\k=1}}^{i-1} a_{ik} \beta_{kj}$   $i = 1,2,...,N$   
 $\forall j = 1,2,3,...,N \quad a_{ij} = \frac{1}{\beta_{ij}} \begin{pmatrix} a_{ij} - \sum_{k=1}^{i-1} a_{ik} \beta_{kj} \end{pmatrix}$   $i = j + 1, j + 2,...,N$ 

In *Block-based* LU decomposition, an *NxN* matrix A can be partitioned into at least 4 matrices:  $A_{11}$  is a *bxb* matrix,  $A_{12}$  is *bx*(*n-b*),  $A_{21}$  is (*n-b*)*xb*, and  $A_{22}$  is (*n-b*)*x*(*n-b*). *b* is the

partition size.  $A = \begin{bmatrix} \frac{A_{11}}{A_{21}} \\ \frac{A_{12}}{A_{22}} \end{bmatrix} = \begin{bmatrix} \frac{L_{11}}{0} \\ \frac{L_{21}}{L_{22}} \end{bmatrix} \cdot \begin{bmatrix} \frac{U_{11}}{0} \\ 0 \end{bmatrix} \cdot \begin{bmatrix} \frac{U_{11}}{0} \\ 0 \end{bmatrix}$ 

#### References

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## Proposed Work

### a Computation

The algorithm is iterative, each *computational* iteration involves: Stage1: Apply Crout's algorithm for finding the LU

decomposition of  $A_{11}$ . Compute the 1<sup>st</sup> row of  $A_{12}$  knowing that  $\alpha_{11} = 1$ , similarly, compute the 1<sup>st</sup> column of  $A_{21}$  knowing that  $\beta_{11} = \alpha_{11}$ .

<u>Stage2</u>: Compute A<sub>12</sub> completely using L<sub>11</sub> from Stage1 to obtain U<sub>12</sub>; similarly, process A<sub>21</sub> completely using U<sub>11</sub> from Stage1 to obtain L<sub>21</sub>. Compute the 1st term of each column of L<sub>21</sub>U<sub>12</sub> for A<sub>new</sub>.

<u>Stage3</u>: Compute A<sub>new</sub> using the results from Stage2. If A<sub>new</sub> is large enough to be partitioned again, Stage1 through 3 will be repeated; otherwise Crout's algorithm is directly applied to A<sub>new</sub> to find its LU decomposition.

b Resource Allocation 1 partition in 1 message

$$nRounds = R = \begin{cases} \left\lfloor \sqrt[N]{b} \right\rfloor + 1 \right] \quad ((N \ mod \ b) + b) > MAX_MX_SIZE \\ \left\lfloor N_{b} \right\rfloor & \text{otherwise} \end{cases}$$

 $nPartitions = P = R^2$   $nNodes = 2 \times R - 1$ 

Distinguish 4 kinds of sensor *nodes* involved in the computation in terms of the type of operations assigned to them:

ClusterHead responsible for the LU operation using Crout's algorithm

- Node with ID2 responsible for lower matrix operation. (even-ID nodes)
  - Grp1 responsible for upper matrix operation. (odd-ID nodes)
  - Grp2 responsible for partitioning/regrouping of the original matrix

Selection algorithm is resource-aware, in terms of power and storage. Three bins for voltage and storage are distinguished.

## ENERGY and STORAGE requirements increase with the iteration number. Nodes in grp2 have higher requirements than those in grp1

- BS initially selects [N/b]<sup>2</sup> nodes out of which only (2\*[N/b]-1) will be active, the rest can sleep and only wakeup during JobDelegation slots
- BS builds a CandidateList for each active node, out of the rest nodes
- CandidateList contains (R-1) CandidateNodes. R = #rounds the node should be active.
- CandidateNodes belong to the same voltage level and same memory level of the active node (within a variance  $\pm \delta$ )





stavs awake

Time slot allocation W<sub>3x3</sub> A<sub>3x3</sub> N<sub>2</sub> N<sub>2</sub> *b*<sub>3x3</sub> N, С<sub>3х4</sub> N, m<sub>3x3</sub> n<sub>3x4</sub> N<sub>5</sub> N<sub>7</sub> x<sub>3x3</sub> N<sub>4</sub> I<sub>3x3</sub> N<sub>2</sub>  $q_{3x3} \\ N_2$  $\begin{array}{c|c} y_{3x3} & o_{3x3} \\ N_6 & N_6 \end{array}$ r 3x4 N<sub>7</sub> 
 36 Slot
 1
 -</  $\begin{array}{c|c} z_{4x3} & p_{4x3} \\ N_8 & N_8 \end{array}$ S<sub>4x3</sub> N<sub>8</sub> virce CH CH vicast Destination N<sub>2</sub> N<sub>2</sub> RX: & R\_+ R\_ Time Slot 1 2 Data Transmitted t Rr RX: L. R., TX: R: nicast Destination N RX: c. R., R., TX: R., R.; nNodes = 2. R -1 = 7 nRounds = R = 4RX: p.R., R., R. RX: a, R., R. TK: R., R., RX: A Roo TX: R  $nMsgs = 2.R^2 + R - 1 = 4$ Detailed operation Experiment Figure1 Setup for the power measurement; the USB attached mote acts as the ter Head, the other motes are distributed around the laboratory. A java GU was built to provide user friendly environment to deal with the motes. 2000 25x25 - 30x30 40x40 P<sub>max</sub> = 4.12x10<sup>-2</sup>W F 1500 Pavo = 4.06x10-2W 6 1000 500 Land-III 4x4 5x5 6x6 7x7 313 3x3 4x4 5x5 6x6 7x7 Figure 2 Total energy (mJ) for different Figure 3 Total time (ms) per operation matrix sizes and different partition sizes for different matrix sizes and different

partition sizes

Illustrative Example

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Applications



and plunger, measuring vibrations and structural properties remotely in a distributed fashion. The distributed, embedded, numerical computational kernel, enables the processing tasks.

A practical

Sensors distributed on lifte



application for vibration measurement on an oil derrick was done as part of project UCOMS (www.ucoms.org)