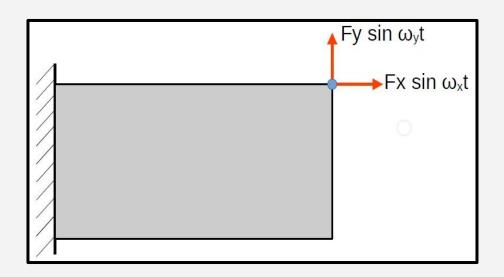


# Parallelization of 0th order Generalised Mode Acceleration Method



Find response in x direction of the node at which force is applied

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# **Algorithm for General Mode Acceleration Method**

Retrieving K and M matrix from existing structural dynamics problem Guyan condensation of M and K matrix from 3782x3782 to 500x500 Solving generalized eigenvalue decomposition of M and K matrix Computing response by 0th order GMAM using **Eigenvectors and Eigenvalues.** Compare the results with MATLAB

# Solving generalized EVP of M and K

EVP of M using QR decomposition

STEP-1:- Initialise Q<sub>o</sub> with Identity matrix
M<sub>o</sub> = M
K<sub>o</sub> = K

STEP-2:- QR decomposition

$$M_k = Q_k * R_k$$

Finding Q<sub>k</sub> using Gram Schmidt Method

Step 3: Calculate M<sub>k+1</sub>

$$\mathbf{M}_{k+1} = \mathbf{Q}^{\mathsf{T}}_{k} * \mathbf{M}_{k} * \mathbf{Q}_{k}$$

STEP-4: Calculating Eigenvalues of M

$$\wedge_{\mathbf{m}} = \mathbf{M}_{\mathbf{n}}$$

STEP-5: Calculating Eigenvectors of M  $\phi_m = Q_n * Q_{n-1} * ... * Q_0$ 

#### contd....

STEP-7 :- Calculating transformation matrix  $\phi(tr)$ 

$$\phi_{\rm tr} = \phi_{\rm m} \star \wedge_{\rm m}^{-1/2}$$

STEP-8 :- Transform K with the help of  $\phi_{tr}$ 

$$\mathbf{K}_{tr} = \boldsymbol{\varphi}_{tr}^{\mathsf{T}} \mathbf{K} \boldsymbol{\varphi}_{tr}$$

STEP-9 : Repeat step 2 to 6 for Eigenvalue decomposition of  $K_{tr}$  matrix

STEP-10 : Calculating eigenvalues and eigenvectors of generalized eigenvalues of K and M

### **MPI Parallelization**

#### **QR decomposition parallelization**

Construction of kth column of Q matrix:-

Formation of Q parallelized by distribution of  $\sum (A_k, Q_n)Q_n$  among the processors.

- Each process has to perform computation of  $\sum (A_k \cdot Q_n)Q_n$  for k/cores columns.
- Root process will compute Temp<sub>K</sub> and Q<sub>k</sub>.

### Further procedures

- Lot of matrix multiplication was involved in algorithm.

All the procedures use Matrix multiplication, which can be parallelised.

## **MPI Parallelization**

## Matrix multiplication parallelization C=AB:-

- Share B matrix with all the processors.
- Share chunks of rows of A with other processors, and remaining rows will be calculated by A.
  - slots=floor(rows of A/ncores), is the chunk of rows being sent to other processors.
- Receive the computed rows of C from other processors and assemble them in C at appropriate positions.

## **CUDA Parallelization**

#### Parallelizable portions of codes are

- 1. Matrix multiplication
- 2. QR decomposition

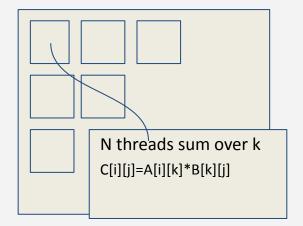
 $r_{n} \dots r_{k-1}$ 

## Matrix multiplication (AB=C) (Each matrix of size N\*N)

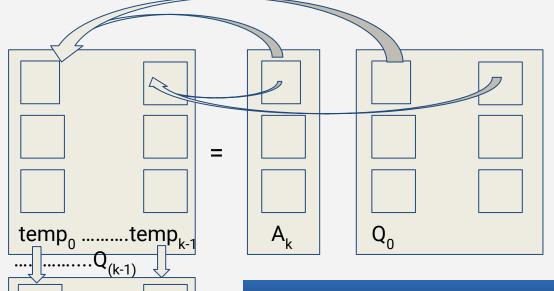
matmult <<< grid(N,N), N >>>(A,B,C)

- This kernel launches NxN blocks with N threads in each block
- Each block calculates one entry of C matrix C[i][j]

# QR decomposition (for calculating $k_{th}$ column of $Q_k$ )



#### Step 1: Calculation of k dot products i.e. $A_k.Q_n$ .



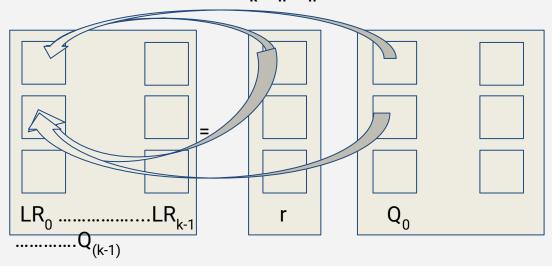
#### dotmat<<< grid(N,N),1 >>>

Each block calculates one entry of temp matrix with single threads

 Performs element wise multiplication

dot<<< 1,N >>> : Each threads add column of temp matrix to get
array of k dot products

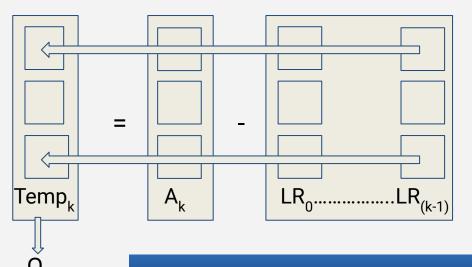
Step 2: Calculate matrix (A<sub>k</sub>.Q<sub>n</sub>)Q<sub>n</sub>



submat<<< grid(N,N),1 >>> Each block calculates one entry of LR matrix with single threads

 Performs element wise multiplication and

Step 3: Calculate Temp<sub>k</sub> and Q<sub>k</sub>



#### Qcal<<< 1,N >>>

- Each threads sums along rows of LR matrix and subtract the sum from corresponding element of A<sub>k</sub>to get element of Temp<sub>k</sub>
- Thread 0 calculated unit vector of Temp<sub>k</sub> i.e. Q<sub>k</sub>

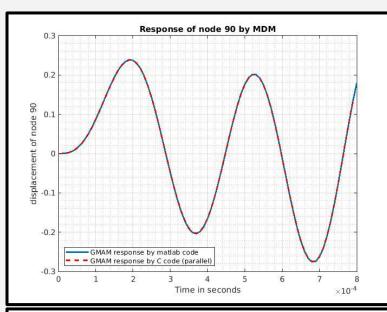
This same process of QR decomposition is done iteratively for n times to calculate whole Q matrix

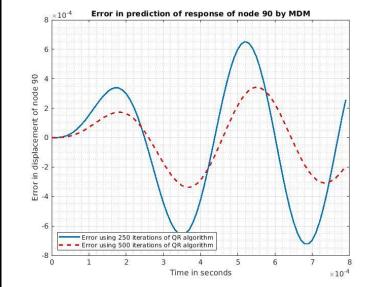
# **Results and discussions**

**Time study**: The serial code took 48 mins to run and below is the time study for different techniques.

OpenMP (min)	OpenMPI (min)	CUDA (min)
8 thds - 9.1	8 PEs - 9.4	8 cores - 11.02
12 thds - 10.4	12 PEs - 9.3	12 cores - 10.9
16 thds - 12.5	16 PEs -10.7	16 cores - 9.7
20 thds - 15.03	20 PEs - 11.1	20 cores - 10.1

- The same problem was solved using Matlab code to confirm that the parallelized code are having data consistency and coherency
- The response of node 3781 (90 in reduced matrix ) is plotted and its error value is also plotted. we can see that the error is of the order  $10^{-4}$  (0.1%)





## **Work Distribution**

- Abhinandan Kumbhar MPI, GPGPU programming, Solving same problem in MATLAB to compare results.
- 2. Jalaj Gupta OpenMP, Solving same problem in MATLAB to compare results.
- 3. R. Nitin Iyer MPI, Guyan condensation
- 4. Saurabh Pal Guyan condensation, GPGPU programming

Everyone has contributed to Report writing, Slides preparation and Serial Code.