# HW<sub>5</sub>

# 1. TASK 1: PING PONG LATENCY

# **Summary:**

I have used 1000 iterations of ping-pong loop to get accurate timings. Finally ran the code for 16 data points with data following message\_length(L) in bytes: 0, 100, 1000, 10000, 100000, 500000, 1000000, 2000000, 3000000, 4000000, 5000000, 6000000, 7000000, 8000000, 9000000, 100000000. The code calculates total timing Tmsg for all the above mentioned data points.

To calculate  $T_s$ , I simply used the value of Tmsg for message\_length(L) =0. To calculate  $T_w$ , I am using mean of  $T_w$  value for all the 16 data points.

 $T_s = 0.007997999 \text{ sec}$  $T_w = 0.00005823312 \text{ sec}$ 

# Loop Iterations used:

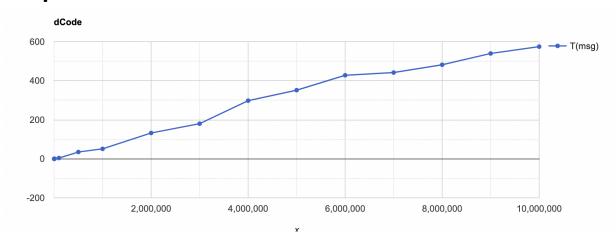
1000

## **Data Points:**

Data Points	X Axis -> Message Length(L)	Y Axis -> T <sub>msg</sub>
1	0	0.007997999
2	100	0.013313002
3	1000	0.048486501
4	10000	0.363008
5	100000	4.0599218
6	500000	34.8613816
7	1000000	51.0772118

8	2000000	132.3538463
9	3000000	179.9873296
10	4000000	297.8985994
11	5000000	351.6244534
12	6000000	427.94704218
13	7000000	441.8691686
14	8000000	481.82289031
15	9000000	539.689148131
16	10000000	574.63355542

# **Graph:**



NOTE: X axis denotes Message length(L) in bytes, and Y axis denotes  $T_{\mbox{\tiny msg}}$  in seconds

# Calculation for T<sub>s</sub>:

- For L=0,  $T_{msg} = T_s$
- From above data points,  $T_{msg}$  for L=0 is 0.007997999 sec
- Therefore,  $T_s = 0.007997999$  sec

# Calculation for T<sub>w</sub>:

- For any data point (L,  $T_{msg}$ )

$$T_{msg} = T_s + T_w * L$$
$$=> T_w = (T_{msg} - T_s)/L$$

Data Points	X Axis -> Message Length(L)	Y Axis -> T <sub>msg</sub>	(T <sub>msg</sub> - T <sub>s</sub> )/L
1	0	0.007997999	
2	100	0.013313002	0.00005315002000000007 10
3	1000	0.048486501	0.00004048850099999982 79
4	10000	0.363008	0.0000355009999999999 97
5	100000	4.0599218	0.00004051923799999984 07
6	500000	34.8613816	0.00006970676720000000 78
7	1000000	51.0772118	0.00005106921379999996 89
8	2000000	132.3538463	0.00006617292415000002 06
9	3000000	179.9873296	0.00005999311053333331 51
10	4000000	297.8985994	0.00007447265034999998 89
11	5000000	351.6244534	0.00007032329
12	6000000	427.94704218	0.00007132317
13	7000000	441.8691686	0.00006312302437142862 4
14	8000000	481.82289031	0.00006022686
15	9000000	539.689148131	0.00005996457223677777 879
16	10000000	574.63355542	0.00005746255574200000 90
SUM -> $\sum_{n=1}^{15} (T_{msg} - T_s)/L$			0.00087349689

- Now, we will take mean of all the values for all data points m to get most accurate value for  $T_{\rm w}$ , i,e,

$$T_{w} = \left(\sum_{n=1}^{m} (T_{msg} - T_{s})/L\right) \div m$$

$$T_{w} = \left(\sum_{n=1}^{15} (T_{msg} - T_{s})/L\right) \div m$$

$$T_{w} = \left(\sum_{n=1}^{15} (T_{msg} - T_{s})/L\right) \div 15$$

$$T_{w} = \left(\sum_{n=1}^{m} (T_{msg} - T_{s})/L\right) \div 15$$

$$T_{w} = 0.00087349689 \div 15$$

$$T_{w} = 0.00005823312 \sec$$

# 2. TASK 2: FINITE-DIFFERENCE FOR 2D DATA DECOMPOSITION

#### **Assume**

- **Dimentions** =  $Nx \times Ny \times Nz$ , where Nx = Ny = N for simplicity
- 9-point stencil
- Number of processes = P
- 2-D domain decomposition

#### Given the dimensions:

Total grid points = 
$$N \times N \times N_z = N^2 N_z$$
 - (1)

## **Calculations:**

#### A. Computation time

Assuming No replicated computation:

$$Tcomp = t_c \times N^2 \times N_Z \qquad -(2)$$

where tc = average computation time per grid point (slightly different at edges from interior etc)

#### **B.** Communication time

Given that its a 2-D domain decomposition, we will partition the grid points among P processors in 2 directions, namely, X and Y.

Assuming that the partitioning is equal in both the directions, **the dimension of task per processor** would then be:

X direction: 
$$\frac{N}{\sqrt{P}}$$
, Y direction:  $\frac{N}{\sqrt{P}}$ , Z direction:  $N_Z$ 

Total grid points per processor = 
$$\frac{N}{\sqrt{p}} \times \frac{N}{\sqrt{p}} \times N_Z$$
 - (3)

Given that its a 9 point stencil and 2d decomposition, each task exchanges 2 planes with each of 4 neighbours.

Size of walls with these 4 neigbours is:

$$\frac{N}{\sqrt{P}} \times N_Z$$

Therefore:

$$T_{comm} = 4P * (T_S + 2 * T_W * \frac{N}{\sqrt{P}} * N_Z)$$
 - (4)

#### C. 2D Finite difference time

Assuming that P divides N exactly, then assume load-balanced and no idle time:

## D. 2D Finite difference Efficiency

We know, relative efficiency is given by:

$$E_{relative} = \frac{T_1}{PT_p}$$
 - (6)

Where,  $T_1$  is time to run algorithm by a single processor, and  $T_p$  is time to run algorithm on P processors

For our algorithm,

$$T_1 = t_c N^2 N_Z \text{ And } T_p = T_{2D\_finite\_diff} = \frac{t_c N^2 N_Z}{P} + 4T_S + 8T_W * \frac{N}{\sqrt{P}} * N_Z)$$

$$\Rightarrow E = \frac{T_1}{PT_p} = \frac{t_c N^2 N_Z}{P^* (\frac{t_c \times N^2 \times N_Z}{P} + 4T_S + 8T_W^* \frac{N}{\sqrt{P}} * N_Z)}$$

$$\Rightarrow E = \frac{t_c N^2 N_Z}{t_c N^2 N_Z + 4PT_S + 8\sqrt{P} T_W N N_Z} - (7)$$

## E. 2D Finite difference Iso-Efficiency

For 2D decomposition, efficiency we calculated in equation 7 is:

$$\Rightarrow E = \frac{t_c N^2 N_Z}{t_c N^2 N_Z + 4PT_S + 8\sqrt{P} T_W N N_Z)}$$

$$\Rightarrow t_c N^2 N_Z = E * (t_c N^2 N_Z + 4PT_S + 8\sqrt{P} T_W N N_Z)$$

Here, dominant terms are:

$$\sim N^2 N_Z \approx \sim N^2 N_Z \sim \sqrt{P} N N_Z$$

For E to be constant,

$$N^{2}N_{z} = \sqrt{P} N N_{z}$$

$$\Rightarrow N = \sqrt{P}$$
- (8)

By substituting  $N = \sqrt{P}$ , E is constant given by equation

$$t_c N_Z = E * (t_c N_Z + 4T_S + 8T_W N_Z)$$
 - (9)

Total computation Tcomp, given by equation 2 is  $Tcomp = t_c N^2 N_Z$ 

Therefore, to keep E constant:  $Tcomp \propto N^2$ 

$$\Rightarrow Tcomp \propto P \ (Using \ Eq \ 8: \ N = \sqrt{P}) \qquad \qquad - \ (10)$$

Number of grid points, given by equation 1 is  $Grid\ points = N^2N_Z$ 

Therefore, to keep E constant:  $Grid\ points \propto N^2$ 

$$\Rightarrow$$
 Grid points  $\propto P$  (Using Eq 8:  $N = \sqrt{P}$ ) - (11)

## From equations 10 and 11, we can conclude:

the isoefficiency of this algorithm is O(P)

$$\Rightarrow$$
 Isoefficiency =  $O(P)$  - (12)

## F. Analysis

## Comparative Analysis of 2D\_Decomposition and 1D\_Decomposition

- 2D\_Decomposition with *Isoefficiency* of O(P) is way more scalable than 1D Decomposition which has *Isoefficiency* of O(P<sup>2</sup>).
- This is because the amount of computation needed to keep E constant in 2D\_Decomposition must increase linearly with the number of processors to keep the efficiency constant. This is way better than 1D\_Decomposition, where the amount of computation needed to keep E constant must increase as the square of the number of processors.

## Reasoning:

- The number of neighbors in 2D\_Decomposition taking part in the communication is 4 per processor, 2 more than the number of neighbors in 1D\_Decomposition per processor.
- Although the amount of communication increases from 2P to 4P, the number of messages/grid\_points shared per communication reduces.
- This is because the wall area is lesser for 2D\_Decomposition( $\frac{N}{\sqrt{P}}*N_Z$ ) in comparison to 1D\_Decomposition's wall area( $N*N_Z$ ).

#### Conclusion:

 Decomposition that increases the number of walls, will end in better efficiency. This is because, although the amount of communication may increase, the number of messages/grid points shared per communication reduces.