

An Algorithm for Optimized Cost in a Distributed Computing System

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Abstract -A distributed system consists of a collection of autonomous computers, connected through a network which enables computers to coordinate their activities and to share the resources of the system. In distributed computing, a single problem is divided into many parts, and each part is solved by different computers. As long as the computers are networked, they can communicate with each other to solve the problem. DCS consists of multiple software components that are on multiple computers, but run as a single system. The ultimate goal of distributed computing is to maximize performance in a time effective, cost-effective, and reliability effective manner. The problem is addressed of assigning a task to a distributed computing system. The assignment of the modules of tasks is done statically. In this paper We study and formulate an algorithm to solve the problem of static task assignment in DCS, each task be assigned to get the more reliable results in lesser cost. The method uses mathematical algorithm by using optimization for optimized cost for task allocation in DCS in MATLAB

Keywords: Distributed Network, Dynamic Allocation, Performance, Residing cost, Reallocation cost

1. INTRODUCTION

Distributed Computing System (DCS) is a collection of independent computers interconnected by transmission channels that appear to the users of the system as a single computer. Distributed systems are groups of networked computers. The word distributed in terms such as DCS, referred to computer networks where individual computers were physically distributed within some geographical area. The terms are Now days used in a much wider sense. Each node of DCS is equipped with a processor, a local memory, and interfaces. The purpose of the distributed system is to coordinate the use of shared resources or provide communication services to the users. In distributed computing, each processor has its own private memory (distributed memory). The processors in a typical distributed system run concurrently in parallel. The required processing power for task assignment applications in a DCS can not be achieved with a single processor. One approach to this problem is to use (DCS) that concurrently process an application program by using multiple processors. as a means of differentiating between the various components of a project. It can also be understand as usually assigned piece of work to the processor often to be finished within a certain time. A task is a piece of code that is to be executed and task

allocation refers to the way that tasks are chosen, assigned, coordinated. Execution time is the time in which single Instruction is executed. Execution cost can be termed as the amount of value of resource used. The execution cost of a task depends on the processor on which it is executed (heterogeneous processors) and the communication between two tasks depends only on whether or not they are assigned to the same processor (homogeneous network). several issues arise such as the minimization of time and cost as well as maximization of system reliability count. By considering that the preference should be given to the idle processor we assign load count as 1 or 0. Now, in each table we will do the addition of each row and will also take the average of each row on the basis of sum of each row. Now, we will subtract the values from average. Negative and zero values will not be considered. For time and cost minimum value will be allocated and for reliability maximum value will be considered. Now the tasks can be allocated for getting the optimized results in terms of cost, also E cost can be calculated. The function for obtaining the overall assignment time, cost and reliability is as follows-Distributed computing is a field of computer science that studies distributed systems. A distributed system is a model in which components located on networked computers communicate and coordinate their actions by passing messages. The components interact with each other in order to achieve a common goal. Three significant characteristics of distributed systems are: concurrency of components, lack of a global clock, and independent failure of components. Examples of distributed system from systems to peer-to-peer applications.

2. Objective

The objective of this research is to find out the optimized cost in Distributed Computing System (DCS) for a task allocation problem or develop a task allocation model that can minimize the overall system cost with the dynamic re-allocation approach. task execution must be completely done before another task takes control of the processor, and the processor environment is homogeneous. This means that the processors have same speeds or processing capabilities. This study offers a mathematical model that allocates the tasks dynamically as tasks executes in various phases. During the particular task execution rest of other task are residing in the particular phase. Execution cost for each phase [EC], inter task communication cost [ITCC], residence cost [RC] of each task on different processors and relocation cost [REC]

for each task are considered to design a dynamic tasks allocation model. To achieve cost optimization in DCS allocation method finds an allocation with minimum allocation cost.

3. Technique

This research considers a distributed computing system consisting of a set $T = \{t_1, t_2, t_3, t_4, \dots, t_m\}$ of m tasks to be allocated on a set $P = \{p_1, p_2, p_3, \dots, p_n\}$ of n processors divided into k phases with criteria tasks m are more than the number of processors n ($m > n$). Execution cost for phase wise of each processor is given in the form of Execution Cost Matrix $ECM(.,.)$ of order $k \times m \times n$. The Residing Cost for residing the unexecuted tasks on the processor is mentioned in Residing Cost Matrix $RCM(.,.)$ of order $k \times m \times n$. The Inter Task Communication Cost between executing and non-executing tasks are also considered and is mentioned in the Inter Task Communication Cost Matrix $ITCCM(.,.)$ of order $m \times k$ and during the processing a task is re-allocate from one processor to another processor then it also obtained some cost i.e. reallocation cost and it is given in the Reallocation Cost.

Matrix $RECM(.,.)$ of order $m \times k$. To calculate $ERCM(.,.)$ for each phase sum up the values of $ECM(.,.)$ and $RCM(.,.)$. Compute the average of each row of $ERCM(.,.)$ and arrange the values in increasing order in $AVG_ROW\ asc()$. Now it selects first n number tasks from $AVG_ROW\ asc()$ and store them in $ERCM_I(.,.)$ and remaining n number of tasks in $ERCM_II(.,.)$ by partitioning $ERCM(.,.)$ into two sub problems. Follow the same process for next n or less than n and solve them using assignment method. Evaluate the Execution Cost, Communication Cost and Reallocation Cost. Follow the same process for all phases and at the end calculate the value of Execution Cost, Communication Cost and Reallocation Cost to obtain the phase wise total execution cost. Calculate the sum of optimal cost of each phase to evaluate the overall optimal cost of distributed computing system.

4. Proposed Method

Step 1: Start Algorithm

Step 2: Input number of tasks as m

Step 3: input the number of processors in n

Step 4: Input the number of phases in k

Step 5: input the Execution Cost Matrix $ECM(.,.)$ of order $k \times m \times n$

Step 6: input the Residing Cost Matrix $RCM(.,.)$ of order $k \times m \times n$

Step 7: input the Inter Task Communication Cost Matrix $ITCCM(.,.)$ of order $m \times k$

Step 8: input the Reallocation Cost Matrix $RECM(.,.)$ of order $m \times k$

Step 9: For $I = 1$ to m

i. $K = I$:

ii. Add the values for $ECM(.,.)$ and $RCM(.,.)$ and store the results in $ERCM(.,.)$

iii. Store the average of each row of $ERCM(.,.)$ and store it in $AVG_ROW()$

iv. sort $AVG_ROW()$ in ascending and store the results in $AVG_Row\ asc()$

v. While (All tasks of $AVG_Row\ asc()$!=SELECTED)

{

a. Make partition of $ERCM(.,.)$ for n tasks, store it in $ERCM_I(.,.)$ and $ERCM_II(.,.)$

b. Apply assignment method on $ERCM_I(.,.)$ and $ERCM_II(.,.)$

}

v. Compute Execution Cost (EC), Inter Task Communication Cost (ITCC) and Reallocation Cost (RC)

vi. Total Cost = EC + ITCC + RC

$I = I + 1$

10. Applying evolutionary optimization technique

vii. Optimal Cost = (Total Cost)

Step 10: State Results

5. Implementation

This research considers a distributed computing system which is made up of four tasks $\{t_1, t_2, t_3, t_4\}$ to be allocated on two processors $\{p_1, p_2\}$ in five phases. The phase wise execution cost of individual processor is known in the form of Execution Cost Matrix $ECM(.,.)$ of order $k \times m \times n$ where k is the number of phases, m is the number of tasks and n is the number of processors. Residing costs for the remaining tasks, except for the executing task, on each processor is also known and mentioned in Residing Cost Matrix $RCM(.,.)$ of order $k \times m \times n$. Inter Task Communication Cost between the executing task and all other task if they are on different processors also taken into the consideration and mentioned in Inter Task Communication Cost Matrix $ITCCM(.,.)$ of order $m \times k$. During the execution an allocated task is shifted from one processor to another processor during the next phase then some cost is incurred in reassignment process at the end of each phase and it is known as reallocation cost. Reallocation cost for the given example is also known and it is mentioned in Reallocation Cost Matrix $RECM(.,.)$ of order $m \times k$.

5.1 Example

This research paper considers a distributed network system which is made up of four tasks $\{t_1, t_2, t_3, t_4\}$ to be allocated on two processors $\{p_1, p_2\}$ in five phases. The phase wise efficiency of individual processor is known in the form of Execution Cost Matrix $ECM(.,.)$ of order $k \times m \times n$ where k is the number of phases, m is the number of tasks and n is the number of processors. The $ECM(.,.)$ is as mentioned here: Residing costs for the remaining tasks, except for the executing task, on each processor is also known and mentioned in Residing Cost Matrix $RCM(.,.)$ of order $k \times m \times n$. $RCM(.,.)$ mentioned here:

ECM (,) =	Phase	Task	Execution Cost	
			p1	p2
	1	t1	5	4
		t2	-	-
		t3	-	-
		t4	-	-
	2	t1	-	-
		t2	7	6
		t3	-	-
		t4	-	-
	3	t1	-	-
		t2	-	-
		t3	3	5
		t4	-	-
4	t1	-	-	
	t2	-	-	
	t3	-	-	
	t4	4	#	
5	t1	5	6	
	t2	-	-	
	t3	-	-	
	t4	-	-	

Inter Task Communication Cost between the executing task and all other task if they are on different processors also taken into the consideration and mentioned in Inter Task Communication Cost Matrix ITCCM(.) or order m x k. ITCCM is mentioned in form of Table 1:

Table 1: Inter Task Communication Matrix

ITCCM(.) =	Tasks ↓	Phases				
		1	2	3	4	5
	t1	-	4	5	3	-
	t2	2	-	4	5	2
	t3	5	3	-	6	3
	t4	2	3	4	-	0

During the execution an allocated task is shifted from one processor to another processor during the next phase then some cost is incurred in reassignment process at the end of each phase and it is known as reallocation cost. Reallocation cost for the given example is also known and it is mentioned in Reallocation Cost Matrix RECM(.) of order m x k. RECM(.) is shown in Table 2:

RCM (,) =	Phase	Task	Residing Cost	
			p1	p2
	1	t1	-	-
		t2	2	3
		t3	3	2
		t4	4	3
	2	t1	2	3
		t2	-	-
		t3	3	4
		t4	2	5

3	t1	4	2
	t2	3	4
	t3	-	-
	t4	4	2
4	t1	2	4
	t2	3	2
	t3	2	3
	t4	-	-
5	t1	-	-
	t2	3	2
	t3	2	3
	t4	2	4

Table 2: Reallocation Cost Matrix

RECM(.) =	Tasks ↓	Phases				
		1	2	3	4	5
	t1	2	2	5	3	-
	t2	3	3	4	4	-
	t3	4	3	3	3	-
	t4	2	4	2	2	-

As per the execution table (ECM), task t1 will execute in phase 1 while remaining tasks i.e. t2, t3 and t4 will be treat as a residing tasks. On the sum up of ECM(.) and RCM(.) will drive another matrix as ERCM:

ERCM (,) =		p1	p2
		t1	5
t2	2	3	
t3	3	2	
t4	4	3	

On evaluating the average of each row of ERCM(.) an linear array avg_row() is obtained here:

$$avg_row() = \begin{Bmatrix} t1 & t2 & t3 & t4 \\ 4.5 & 2.5 & 2.5 & 3.5 \end{Bmatrix}$$

On arranging the values of avg_row() in ascending order a new linear array avg_row_asc() linear is formed:

$$avg_row_asc() = \begin{Bmatrix} t2 & t3 & t4 & t1 \\ 2.5 & 2.5 & 3.5 & 4.5 \end{Bmatrix}$$

In order to get optimal assignment, allocation technique will divide ERCM(.) by selecting first two tasks from sum_row_asc() and store the values in ERCM_I(.) and last two tasks into ERCM_II respectively:

ERCM_I(.) =		p1	p2
		t2	2
t3	3	2	

ERCM_II(.) =		p1	p2
		t4	4
t1	5	4	

On derived two matrices i.e. ERCM_I and ERCM_II apply assignment method to allocate the tasks and the allocation and their costs is present in Table 3.

Phase	Task	Processor	Assigned Task	Execution Cost (EC)	Communication Cost (CC)	Reallocation Cost (RC)	Phase-wise Total Cost (EC + CC + RC)
1	t ₁	p ₁	t ₂ * t ₄	6	4	0	
		p ₂	t ₃ * t ₁	6			

By applying the same process on the remaining phases final allocation is obtained as present in Table 4 for given example.

Table 4: Final Dynamic Allocation Table

Phase	Executing Task	Processor	Assigned Task	Phase-wise Total Cost (EC + CC + RC)
1	t ₁	p ₁	t ₂ * t ₄	16
		p ₂	t ₃ * t ₁	
2	t ₂	p ₁	t ₁ * t ₄	22
		p ₂	t ₃ * t ₂	
3	t ₃	p ₁	t ₂ * t ₃	19
		p ₂	t ₁ * t ₄	
4	t ₄	p ₁	t ₄ * t ₃	24
		p ₂	t ₂ * t ₁	
5	t ₁	p ₁	t ₃ * t ₄	15
		p ₂	t ₂ * t ₁	
Total Task scheduling cost				96

6. Conclusion

This research designed a task allocation model with dynamic reallocation technique for execution of tasks in Distributed Computing System (DCS) and provides the optimal solution in order to get optimized costs for task allocation. This allocation model considered the several factors of dynamic environment i.e. execution cost, residing cost, reallocation cost, inter task communication cost and most important execution phases. In dynamic model a tasks completes its execution in various phase so presented dynamic allocation model provide optimal solution phase wise. The presented model is tested in MATLAB platform by creating distributed environment as mentioned in Fig. 1. Optimal cost is calculated for each phase and every task. Communication cost between executing and non-executing task, reallocation cost of task are also added to evaluate final optimal cost of each phase. Phase wise results are generated in MLATLAB for presented algorithm and algorithm [17], results are compared for both algorithms, on comparing phase wise execution cost and total execution cost, it is found presented model shows the better results as mentioned in Table 7

Table 7: Algorithm results derived in MATLAB environment and compare with algorithm [17]

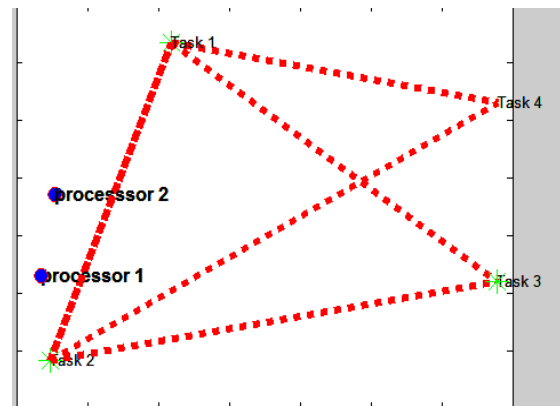


Fig.1: Distributed environment in MATLAB

Overall results also evaluated and compared with past algorithm and found to be very less as shown in Table

Table:5 Comparison Table

	Total cost
Proposed Algorithm	96
Earlier Algorithm	115

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