Performance Upgradation through Task Allocation of Distributed Networks

Kapil Govil

School of Engineering & Technology, ITM University Gwalior

ABSTRACT

Normally the distributed network has to execute the tasks that shall be the more than the number of processors. The assignment problem is a case of linear programming helps to solve the problems related to tasks and processors. The problem of execution of "m" tasks to "n" processors (m > n)in a distributed networks is addressed here through a new modified tasks allocation policy for distributed networks. The model, presented in this paper allocates the tasks or modules to the processor to increase the performance and to reduce the execution time. This paper reduces the problem of allocation of tasks where number of processors is less than the number of tasks. The example mentioned in the paper has three tasks and solved it in such a way that the task t_1 processed with minimum time, the task t_2 with minimum cost while the task t_3 with maximum reliability. In this problem, the tasks are fused (or clubbed) with another task(s) on the basis of minimum communication cost to form a balanced allocation.

Keywords

Allocation, Cost, Distributed Network, Performance, Processor, Reliability, Task, Time.

1. INTRODUCTION

Such type of research problems in which the performance of the distributed systems is to be upgraded, requires either processing time or cost to be minimized or reliability to be maximized by deciding the strategy of allocation of tasks to the processors of the distributed systems. These problems may be categorized as static (15, 16, 17, 19, 29) and dynamic (2, 14, 18, 19, 26, 27) in nature. Some of the other related methods have been reported in the literature, such as, Integer programming (7, 23), Branch and Bound technique (28), Matrix Reduction technique (11, 30, 31), Reliability Optimization (1, 12, 20, 21, 24), Load Balancing (2, 9, 10) and Modeling (3, 6, 8). The series parallel redundancyallocation problem has been studied with different approaches, such as, Dynamic programming (4, 10, 13), Integer programming (7, 23), and Heuristic techniques (5, 22, 25).

2. OBJECTIVE

The objective of the present research paper is to enhance the performance of the distributed systems by using the proper utilization of its processors. A set of tasks have to be processed by the processors of the network, while each of the task have the modules and the number of modules are more than the number of processors of the network. The processing of a task is means that all of its modules get processed. Performance is the measure in term of either time or cost or reliability of the modules of a task that have to process on the processors of the system and these have to be optimally processed i.e., either time or cost to be minimized or reliability to be maximized.

3. TECHNIQUE

To evaluate the optimal time or cost or reliability for each task through optimal allocation, initially it has to concentrate on those (m-n) modules that have the highest probability of data transfer with the remaining n modules. Each of these (m-n)modules (say m_{ik}) of every task is treated as a candidate to be fused with any one (say m_{il}) out of the remaining n modules with which it has the highest communication. Further, all the elements of k^{th} row and l^{th} row are to be added in case of time and cost while in case of reliability these rows have to multiply. This will reduces the effectiveness matrix for each task in to a square matrix. Now the problem remains to determine the optimal time or cost or reliability through the allocation strategy by considering either task processing based on time or cost or reliability for all modules to individual processor(s) for each task. For allocation purpose a modified version of row and column assignment method proposed by Kumar et al (17) is employed which allocates all the modules of a task to a processor optimally. The functions for obtaining the overall assignment execution time [Etime], execution cost [Ecost], and execution reliability [Ereliability] are as follows:

$$Ptime = \left[\sum_{i=1}^{n} \left\{\sum_{j=1}^{n} PT_{ij}x_{ij}\right\}\right]$$
(1)
$$Pcost = \left[\sum_{i=1}^{n} \left\{\sum_{j=1}^{n} EC_{ij}x_{ij}\right\}\right]$$
(2)
$$\Pr elibility = \left[\prod_{i=1}^{n} \left\{\sum_{j=1}^{n} ER_{ij}x_{ij}\right\}\right]$$

(3) Where, $x_{ij} = \begin{cases} 1, \text{ if } i^{th} \text{ task is assigned to } j^{th} \text{ processor} \\ 0, \text{ otherwise} \end{cases}$

4. ALGORITHM

Step 1:	Start algo
Step 2:	Read the number of tasks in m
Step 3:	Read the number of processors in n
Step 4:	For $I = 1$ to n
Step 5:	For $J = 1$ to m
Step 6:	Read the value in
	PTM [I][J]
Step 7:	Increase the value of J by 1
Step 8:	End of J loop
Step 9:	Increase the value of I by 1
Step 10:	End of I loop
Step 11:	For $I = 1$ to n
Step 12:	For $J = 1$ to n

Step 13:	Read the value in CM [I][J]					
Step 14:	Increase the value of J by 1					
Step 15:	End of J loop					
Step 16:	Increase the value of I by 1					
Step 17:	End of I loop					
Step 18:	For $I = 1$ to n					
Step 19:	For $J = I$ to n					
Step 20:	If $CM[I][J] == 1$ then					
Step 21:	Store the value of					
	P1[I] to 1					
Step 22:	Store the value of					
	P2[J] to 1					
Step 23:	Calculate MAT [I][J]					
	= PT/C/RM[P1[I]] + PT/C/RM[P2[J]]					
Step 24:	End of if statement					
Step 25:	Increment the value of J by 1					
Step 26:	End of J loop					
Step 27:	Increase the value of I by 1					
Step 28:	End of I loop					
Step 29:	For $I = 1$ to n					
Step 30:	If $P1[I] == 0$ then					
Step 31:	Store the value of T1[I] by P1[I]					
Step 32:	End of if statement					
Step 33:	Increase the value of I by 1					
Step 34:	End of I loop					
Step 35:	For $I = 1$ to n					
Step 36:	If $T1[I] = 0$ then					
Step 37:	Calculate MAT [T1[I]] =					
	PT/C/RM [T1[I]] +					
	PT/C/RM[T1[I+1]]					
Step 38:	End of if statement					
Step 39:	Increment the value I by 1					
Step 40:	End of I loop					
Step 41:	Count the zero(s) in each row					
Step 42:	Mark the row(s), which have single zero					
Step 43:	Mark the column, which have single zero					
Step 44:	Go to the row(s), which have more than					
	one zero. Now select any one zero and					

	cross the leading zero(s), which are in
	same row and column
45:	Mark the assignments
46:	Count the total assignment
47:	If total number of assignment < order of matrix
48:	Go to Step 52
49:	Else
50:	Go to Step 59
51:	End of if statement
52:	Mark the rows for which assignment have not been made
53:	Mark column that have zeros in marked rows
54:	Mark rows that have assignment in marked column
55:	Repeat Step 53 & Step 54 until chain of marking ends

- Step 56: Draw the minimum number of lines through unmarked rows and marked columns to cover all zeros
- Step 57: Select the smallest element of the uncovered elements and replace it by zero. Also add this element to positions at which lines intersect to each other only

Step Step Step Step Step Step Step Step

Step Step

Step

Step 59: State processing time Step 60: End algo

5. IMPLEMENTATION

Consider an example consisting of a set $T=\{t_1,t_2,t_3\}$ of 3 tasks each of them having sets $M_1{=}\{m_{11},m_{12},m_{13},m_{14},m_{15}\}$ of 5 modules, $M_2{=}\{m_{21},m_{22},m_{23},m_{24}\}$ of 4 modules and $M_3{=}\{m_{31},m_{32},m_{33},m_{34},m_{35},m_{36}\}$ of 6 modules respectively. The three processors are available in the distributed network to process the tasks that are represented by the set $P=\{p_1,p_2,p_3\}$. The processing time (t), cost (c) and reliability (r) of each module of every task on various processors are known and mentioned in the following matrix, namely, PCTR (,):

Step 58: Go to Step 42

[-
	Processors	p_{1}	p_2	p_{3}
Tasks	Modules	t-c-r	t-c-r	t-c-r
t ₁	m_{11}	110 - 2300 - 0.999856	130 - 2400 - 0.999420	090 - 2500 - 0.999635
	m_{12}	110 - 2100 - 0.999632	140 - 2700 - 0.999632	080 - 2800 - 0.999120
	m 13	080 - 2800 - 0.999235	150 - 2800 - 0.999541	150 - 2100 - 0.999631
	m_{14}	070 - 2000 - 0.999632	120 - 3000 - 0.999520	120 - 2700 - 0.999687
	m 15	120 - 2400 - 0.999863	070 - 2300 - 0.999235	110 - 2200 - 0.999631
<i>t</i> 2	m_{21}	170 - 2500 - 0.999652	140 - 2200 - 0.999632	150 - 2300 - 0.999785
	m_{22}	160 - 2100 - 0.999856	150 - 2500 - 0.999652	140 - 2900 - 0.999412
	<i>m</i> ₂₃	180 - 2700 - 0.999523	200 - 2700 - 0.999632	120 - 2300 - 0.999652
	m_{24}	110 - 2400 - 0.999532	100 - 2600 - 0.999632	190 - 2800 - 0.999874
<i>t</i> 3	m 31	110 - 2100 - 0.999452	130 - 1600 - 0.999625	090-1800-0.999856
	<i>m</i> ₃₂	110 - 2900 - 0.999466	140 - 3000 - 0.999653	080 - 2200 - 0.999785
	<i>m</i> ₃₃	080 - 2400 - 0.999256	150 - 2800 - 0.999245	150 - 2200 - 0.999631
	M 34	070-1200-0.999215	120 - 3600 - 0.999123	120 - 2300 - 0.999563
	M 35	120-1800-0.999452	070 - 2000 - 0.999365	110 - 2100 - 0.999887
L	M 36	130 - 2700 - 0.999785	080 - 2000 - 0.999632	100 - 2200 - 0.999562

The communication period amongst the modules of each task has also been considered and it is mentioned in the following matrices, namely, CM(,):

For task t_1 ,	the matrix	CM (1,)) is as:
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_	m_{11}	m_{12}	m 13	m_{14}	m 15]	
m_{11}	0	1	6	9	3		
m 12		0	2	7	8		
m 13			0	4	5	;	
m_{14}				0	2		
m 15					0		
For task t_2 , the matrix CM (2,) is as:							
_ For	task t_2 ,	the ma	trix Cl	M (2,)	is as:		
For	task t_2 , m_2	the main m_1	trix Cl 122 1	M (2,) N ₂₃	is as: m_{24}		
For m_{21}	task t_2 , m_2 0	the matrix \mathcal{M}	utrix CI 122 1 2	M (2,) n ₂₃ 4	is as: m_{24} 5		
For m_{21} m_{22}	task t ₂ , <i>M</i> 2 0	the ma <i>M</i> 2 (utrix Cl 222 <i>P</i> 2)	M (2,) n ²³ 4 7	is as: <i>m</i> ₂₄ 5 8	:	
For m_{21} m_{22} m_{23}	task t ₂ , <i>m</i> 2 0	the ma	atrix Cl 222 7 2	M (2,) n ₂₃ 4 7 0	is as: m_{24} 5 8 6	;	
For m_{21} m_{22} m_{23} m_{24}	task t ₂ , <i>m</i> 2 0	the ma 1 <i>M</i> 2 (utrix CI 222 1)	M (2,) n ₂₃ 4 7 0	is as: m_{24} 5 8 6 0	;	

For task t_3 , the matrix CM (3,) is as:							
	m ₃₁	M 32	M 33	M 34	M 35	M 36	
<i>m</i> ₃₁	0	2	3	7	9	6	
<i>m</i> ₃₂		0	4	8	6	5	
<i>m</i> ₃₃			0	1	5	4	
<i>m</i> ₃₄				0	2	3	
<i>m</i> ₃₅					0	1	
<i>m</i> ₃₆						0	

Here, it is considered the processing of the tasks t_1 based on the time constraints (however one may choose the cost or reliability constraints also); t_2 is based on the cost constraints (however one may choose the time or reliability constraints also); and for the t_3 it is based on reliability constraints (however one may choose the time or cost constraints also). Further it is also noted that each task has modules that are more than the number of processors in the distributed system. So following data from the matrix PCTR (,) is used i.e,

-	Processors	p_{1}	p_2	<i>p</i> ₃
Tasks	Modules	t-c-r	t-c-r	t-c-r
t_1	m_{11}	070 - · · · - · · ·	$110 - \cdots - \cdots$	120
	m_{12}	080	090 - · · · - · · ·	130
	m 13	120	135 - · · · - · · ·	$140 - \cdots - \cdots$
	m_{14}	160	$140 - \cdots - \cdots$	150
	m 15	$170 - \cdots - \cdots$	130	160
t_2	m_{21}	$\cdots - 2500 - \cdots$	$\cdots - 2200 - \cdots$	$\cdots - 2300 - \cdots$
	m 22	$\cdots - 2100 - \cdots$	$\cdots - 2800 - \cdots$	$\cdots - 2900 - \cdots$
	m 23	$\cdots - 2700 - \cdots$	$\cdots - 2700 - \cdots$	$\cdots - 2300 - \cdots$
	m_{24}	$\cdots - 2400 - \cdots$	$\cdots - 2600 - \cdots$	$\cdots - 2800 - \cdots$
t_3	m ₃₁	$\dots - \dots - 0.999452$	0.999625	0.999856
	M 32	0.999466	0.999653	$\dots - \dots - 0.999785$
	<i>m</i> ₃₃	0.999256	$\dots - \dots - 0.999245$	0.999631
	M 34	0.999215	0.999123	0.999563
	M 35	0.999452	0.999365	0.999887
_	M 36	$\dots - \dots - 0.999785$	0.999632	0.999562

The task t_1 has five modules, so that on the basis of highest communication, the modules $m_{11} \& m_{14}$ and $m_{12} \& m_{15}$ are fused together to reduce the effectiveness matrix square. The task t_2 has four modules, so that on the basis of highest communication, the modules $m_{21} \& m_{24}$ are fused together to

reduce the effectiveness matrix square. The task t_3 has six modules, so that on the basis of highest communication, the modules m_{31} & m_{35} , m_{32} & m_{32} and m_{34} & m_{36} are fused together to reduce the effectiveness matrix square. The resulting matrix is as:

The results of the allocations based on cost for the task t_2 are

obtained after implementing the row & column assignment

process as suggested by Kumar et al (17), are mentioned

below in the Table 2;

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	Processors	p_1	p_2	p_3
Tasks	Modules	t-c-r	t-c-r	t-c-r
t ₁	$m_{11*}m_{14}$	230	250	270
	$m_{12} * m_{15}$	250	220	290
	<i>m</i> ₁₃	120	135	$140 - \cdots - \cdots$
t ₂	m_{21}	$\cdots - 2500 - \cdots$	$\cdots - 2200 - \cdots$	$\cdots - 2300 - \cdots$
	$m_{22} * m_{24}$	$\cdots - 4500 - \cdots$	$\cdots - 5400 - \cdots$	$\cdots - 5700 - \cdots$
	<i>m</i> ₂₃	$\cdots - 2700 - \cdots$	$\cdots - 2700 - \cdots$	$\cdots - 2300 - \cdots$
t ₃	$m_{31*}m_{35}$	$\dots - \dots - 0.998904$	0.998990	0.999743
	<i>m</i> ₃₂ • <i>m</i> ₃₄	0.998681	$\dots - \dots - 0.998776$	0.999348
	<i>m</i> ₃₃ • <i>m</i> ₃₆	0.999041	$\dots - \dots - 0.998877$	0.999193

The results of the allocations based on time for the task t_i are obtained after implementing the row & column assignment process as suggested by Kumar et al (17), are mentioned below in the Table 1;

Table 1. Time based Allocation for task t_1			Т	able 2	. Cost based	Allocatio	on for task t_2		
Modules	\rightarrow	Processors	Time	Etime	Modules	\rightarrow	Processors	Cost	Ecost
$m_{11} * m_{14}$	\rightarrow	p_1	230		$m_{22} * m_{24}$	\rightarrow	p_1	2200	
$m_{12} * m_{15}$	\rightarrow	p_2	220	590	m_{21}	\rightarrow	p_2	4500	9000
m_{13}	\rightarrow	p_3	140		<i>m</i> ₂₃	\rightarrow	p_3	2300	

The results of the allocations based on reliability for the task t_3 are obtained after implementing the row & column assignment process as suggested by Kumar et al (17), are mentioned below in the Table 3;

Thus the complete results for the above mentioned example obtained and are mentioned in the Table 4.

Table 4. Optimal Allocation Table

			Processors		Optimal	Optimal	Optimal
Tasks		p_1	p_2	p_{3}	Etime	Ecost	Ereliabili ty
t_1	\rightarrow	$m_{11}*m_{14}$	$m_{12}*m_{15}$	m 13	590		
<i>t</i> 2	\rightarrow	$m_{22}*m_{24}$	m_{21}	<i>m</i> ₂₃		9000	
<i>t</i> 3	\rightarrow	<i>m</i> ₃₁ * <i>m</i> ₃₅	$m_{32}*m_{34}$	<i>m</i> ₃₃ * <i>m</i> ₃₆			0.9966876

6. CONCLUSION

This paper chooses the problem, in which the numbers of module of the tasks are more than the number of processors of the distributed system. The model addressed in this paper is based on the consideration of processing time, cost and reliability of the module of the tasks to the various processors. The communication period amongst the module of the tasks is also used. The method is presented in algorithmic form and implemented on the several sets of input data to test the performance and effectiveness of the algorithm. As it is the common requirement for any assignment that the tasks have to be processed either with minimum time or minimum cost or maximum reliability. The example mentioned in this paper has three tasks and solved it in such a way that the task t_1 processed with minimum time, the task t_2 with minimum cost while the task t_3 with maximum reliability. The optimal results are mentioned in Table 4 of the previous section. The Table 5(a), 5(b) and 5(c) shows the optimal results as obtain after implementing the present algorithm for all three options viz. time, cost, and reliability for each and every task.

Table 5(a). Optimal results for task t₁

Processor	Assignment based on				
	time	Cost	reliability		
p ₁	$m_{11} * m_{14}$	m ₁₃	m ₁₃		
p ₂	$m_{12} * m_{15}$	$m_{11} * m_{14}$	$m_{11} * m_{15}$		
p ₃	m ₁₃	$m_{12} * m_{15}$	$m_{12} * m_{15}$		
Optimal Result	590	12000	0.996929		

Table 5(b). Optimal results for task t₂

Processor	Assignment based on		
	time	Cost	reliability
p ₁	m ₂₁	m ₂₂ * m ₂₄	m ₂₃
p ₂	m ₂₂ * m ₂₄	m ₂₁	m ₂₁
p ₃	m ₂₃	m ₂₃	$m_{22} * m_{24}$
Optimal Result	540	9000	0.998442

Table 5(c).	Optimal	results	for	task	t ₃
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Processor	Assignment based on		
	time	cost	reliability
p ₁	m ₃₁ * m ₃₅	$m_{32} * m_{34}$	$m_{31} * m_{35}$
p ₂	m33 * m36	$m_{31} * m_{35}$	$m_{32} * m_{34}$
p ₃	m ₃₂ * m ₃₄	$m_{33} * m_{36}$	$m_{33} * m_{36}$
Optimal Result	610	12100	0.9966876

Table 3. Reliability based Allocation for task t₃

Modules	\rightarrow	Processors	Reliabilit y	Ereliab ility
$m_{31} * m_{35}$	\rightarrow	p_1	0.998904	
$m_{32} * m_{34}$	\rightarrow	p_2	0.998776	0.9966876
$m_{33} * m_{36}$	\rightarrow	p_3	0.999193	

7. TIME COMPLEXITY

It is known that the analysis of an algorithm is mainly focuses on time complexity. Time complexity is a function of input size 'n'. It is referred to as the amount of time required by an algorithm to run to completion. The time complexity of the above mentioned algorithm is O (m²n²). By taking several input examples, the above algorithm returns results as mentioned in Table 6.

No. of No. of tasks Optimal				
processors	(m)	Results		
(n)				
3	4	144		
3	5	225		
3	6	324		
3	7	441		
3	8	576		
4	5	400		
4	6	576		
4	7	784		
4	8	1024		
4	9	1296		
5	6	900		
5	7	1225		
5	8	1600		
5	9	2025		
5	10	2500		

The graphical representations of the results are shown by Fig 1, 2 and 3.



Fig 1: Graphical representation of results where n=3

Table 6. Time Complexity



Fig 2: Graphical representation of results where n=4



Fig 3: Graphical representation of results where n=5

8. COMPLEXITY COMPARISION

The performance of the algorithm is compared with the algorithm suggested by Richard et al (28). Following Table 7 shows the time complexity comparison between algorithm (28) with present algorithm.

Processors	Tasks	Time	Time
n	m	Complexity	Complexity
		of	of present
		algorithm	algorithm
		$(28) O(n^m)$	$O(m^2n^2)$
3	4	81	144
3	5	243	225
3	6	729	324
3	7	2187	441
3	8	6561	576
4	5	1024	400
4	6	4096	576
4	7	16384	784
4	8	65536	1024
4	9	262144	1296
5	6	15625	900
5	7	78125	1225
5	8	390625	1600
5	9	1953125	2025
5	10	9765625	2500

Table 7. Complexity Comparison

From the Table 7 it is clear that present algorithm is much better for optimal allocation of tasks that upgrade the performance of distributed system. The graphical representation of the data as mentioned in Table 7 is shown through Fig 4, 5 and 6.



Fig 4: Graphical representation of data where n=3







Fig 6: Graphical representation of data where n=5

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