

## 1. INTRODUCTION

Large-scale distributed computing systems (LDCS) coordinate resource sharing in dynamic environments and have emerged as a popular computing platform for computationally intensive jobs.

However, diverse characteristics in both jobs and resources impose a great burden on resource allocation.

We present a dynamic job-clustering technique that enables judicious resource allocation decisions achieving better performance and utilization.

### Objectives:

- Minimize response time
- Maximize resource utilization

## 2. PRELIMINARIES

- Each job  $i$  has  $[w, d]$ , job weight ratio, ( $w/d$ ), and deadline factor  $df_i$ .
- 3 types of job:
  - (i)  $w \leq d$ , (ii)  $w \approx d$  or (iii)  $w \geq d$
- Each site  $x$  has various compute nodes and each node  $j$  is characterized by  $[p, a]$
- Global scheduler acts as decision maker for  $i$ , required to be map into  $j$ .
- It may not be considered task-mapping if the availability of  $j$  is very low (i.e., overloaded)

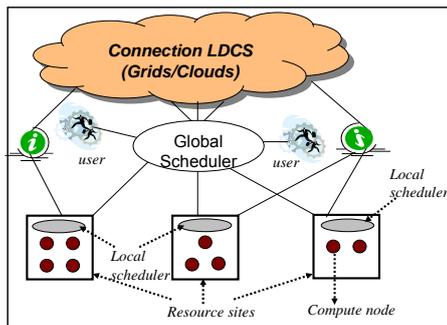


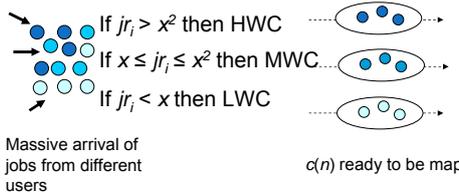
Figure 1. System Model

## 3. METHODOLOGY

- To address the problem of allocating resources to jobs in a dynamic fashion.
- 2 heuristic methods are involved :
  - (i) job-clustering and
  - (ii) mapping scheme.
- Integrate clustering and mapping into resource allocation decisions.
- Help organize mapping alternatives.
- Improve performance and availability

### (i) Job clustering

Jobs group by  $jr_i$  and priority level,  $p_i$  [1]. Given,  $x$  and  $x^2$  are low  $p_i$  and high  $p_i$ , respectively. The jobs been grouped called  $c(n)$  where  $n$  is cluster name (i.e., Heavy, Medium or Light Weight Cluster).



Massive arrival of jobs from different users

$c(n)$  ready to be map

### Each $c(n)$ has :

- Different total jobs,  $val_{c(n)}$ .
- Total  $jr_i$  and total  $w_i$  based on value of  $val_{c(n)}$ .
- Deadline factor  $df_{c(n)}$  [1] from normalized deadline  $nd_{c(n)}$ .
- Closeness rate  $crate_{c(n)}$ ; denotes 1/total differences of  $jr_i$  in  $c(n)$ .  
(If  $crate_{c(n)} \geq 0.1$  then high correlation; otherwise its low correlation).

Table 1. Job features in each  $c(n)$

Cluster Name	Computational size of each job, $w_i$ .	Deadline of each job, $d_i$ .	Deadline factor, $df_{c(n)}$ .
LWC	Small	Long	$\leq 0.25$
MWC	Medium	Moderate	$\leq 0.50$
HWC	Large	Short	$\leq 1.00$

### (ii) Task-resource mapping

Each  $c(n)$  associated with its computational worth; given as total  $jr_i \times df_{c(n)}$ . The task needs to be mapped onto an Available node regardless of site autonomy.

If (waiting time  $\leq$  average execution time) then:  
Resource is set to Available  
Else  
Resource is set to Not\_Available

For mapping event, the fitness value is used that defined as worth/ETC, where ETC = (total  $w_i / p_i$ ).

If  $crate_{c(n)}$  is high correlation then maps to  $j$  that gives highest fitness value  
If  $crate_{c(n)}$  is low correlation then maps to  $j$  that gives second highest fitness value

Advantage :

No reservation technique applied due to no preference is imposed for high, medium and low priority  $c(n)$  during mapping process.

## 4. RESULTS

### Response Time

- It is observed that response time using our strategy (*Fit-Clustering*) is faster than other schemes; in [2] and [3].
- As we used fitness value; even if  $c(n)$  has low worth it may still be picked first if ETC is smaller.

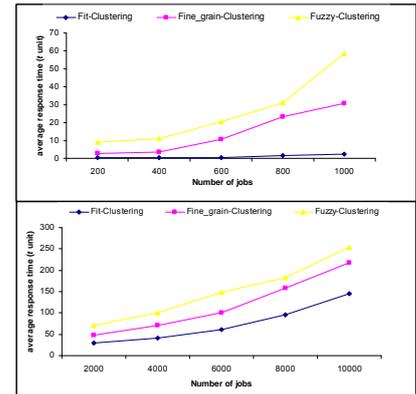


Figure 2. Average response time in moderate and large number of jobs

### Utilization Rate

- In utilization rate, *Fit-Clustering* performs better, especially in large number of jobs by 10% on average.

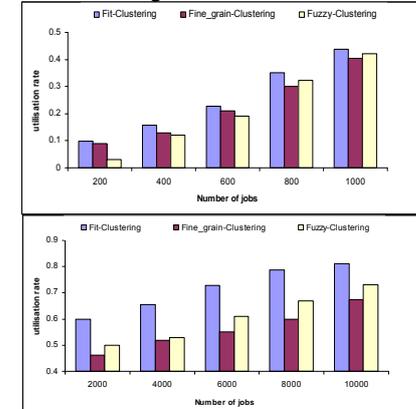


Figure 3. Utilization rate in moderate and large number of jobs

## 5. CONCLUSION

- We addressed the adaptive resource allocation problem with dynamically group of independent jobs; to paves the way in making LDCS efficiently usable for various types of jobs.
- Our clustering strategy effectively exploits both the diversity of jobs and heterogeneity of resources.
- Apparently, well-understandable properties in each group of jobs are important and worthy to investigate for judicious allocation.

### REFERENCES

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