# Weighted Clustering Based Preemptive Scheduling For Real Time System

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*Abstract-* In this paper a new improved clustering based scheduling algorithm for a single processor environment is proposed. In the proposed method, processes are organized into non-overlapping clusters. For each process the variance from the median, is calculated and compared with the variance from the means of other clusters. Each process is assigned to the cluster associated with the closest median. The new median of each cluster is calculated and the procedure is repeated until the medians are fixed. Weight is assigned to each cluster using the externally assigned priorities and the burst time. The cluster with highest weight is executed first and jobs are scheduled using the Round Robin algorithm with calculated dynamic time slice.. The experimental study of the proposed scheduling algorithm shows that the high priority jobs can be executed first to meet the deadlines and also prevents starvation of processes at the same time which is crucial in a real time system.

# Keywords: K-Median Clustering, Service Time, Weight, Round Robin, Variance

# I. INTRODUCTION

Scheduling is a key concept in operating system design. In a multiprogramming environment, the processes that are loaded into the memory compete for processor time [1]. Scheduling determines which process will progress and which will wait. The main objectives that scheduling function must satisfy include fairness, efficient use of processor time, response time, turnaround and throughput Operating systems may be feature up to three distinct types of scheduling: long-term scheduling, medium-term scheduling and short-term scheduling. CPU scheduling is an essential operating system task; therefore its scheduling is central to operating system design. When there is more than one process in the ready queue waiting its turn to be assigned to the CPU, the operating system must decide through the scheduler, the order of execution. Many scheduling algorithms have been proposed for processor assignment, but there is no suitable algorithm for all purposes. So special care is needed to choose and coordinate parameters, e.g. waiting time, total time and utilization. Some of the commonly used algorithms are First Come First Serve(FCFS), Round Robin (RR), Shortest Job First (SJF), Shortest Remaining Time (SRT), High Response Ratio Next (HRRN) and Feed Back (FB) algorithm[3]. Nonpreemptive algorithms like FCFS and SJF are not suitable for real time system, but preemptive scheduling like MLFQ and RR are used to provide good response time and fair dispatching of CPU time e.g. in interactive systems[3]. Clustering of processes is nothing but grouping of the processes into classes or clusters, so that processes within a cluster have high similarity in comparison to one another, but are very dissimilar to processes in other clusters. Cluster analysis is an exploratory data analysis tool which aims at sorting different processes into groups in a way that the degree of association between two processes is maximal if they belong to the same group and minimal otherwise. To achieve different application purposes, a large number of clustering algorithms have been developed. Clustering techniques can be broadly divided into five classes: hierarchical methods, partitioning methods, density-based methods, grid based method, model based method Partitioning clustering algorithms, such as K-means, K-medoids, PAM which assign objects into k (predefined cluster number) clusters, and iteratively reallocate objects to improve the quality of clustering results [8]. Hierarchical clustering algorithms assign objects in tree-structured clusters, i.e., a cluster can have data points or representatives of low level clusters. Hierarchical clustering algorithms can be classified into categories according their clustering process: agglomerative and divisive [9].Density-based clustering is that for each instances of a cluster the neighborhood of a given radious has to contain at least a minimum number of instances. Grid-based clustering first quantize the clustering space into a finite number of cells.cells that contain more than certain number of points are treared as dense and the dense cells are connected to form the clusters. In model based method in addition to the observed or predictive attributes, there is a hidden variable which reflects the cluster membership for every case in the data set. One simple rule of thumb sets the number to  $k \approx \sqrt{n/2}$  with *n* as the number of objects (data points).

In statistics and machine learning, *k*-medians clustering is a variation of *k*-means clustering where instead of calculating the mean for each cluster to determine its centroid, one instead calculates the median. This has the effect of minimizing error over all clusters with respect to the 1-norm distance metric, as opposed to the square of the 2-norm distance metric (which *k*-means does). This relates directly to the *k*-median problem which is the problem of finding *k* centers such that the clusters formed by them are the most compact. Formally, given a set of data points *x*, the *k* centers  $c_i$  are to be chosen so as to minimize the sum of the distances from each *x* to the nearest  $c_i$ . The criterion function formulated in this way is sometimes a better criterion than that used in the *k*-means clustering algorithm, in which the sum of the squared distances is used.

#### II. RELATED WORKS

Yaashuwanth C & R.Ramesh[5]proposed an architecture which eliminates the defects of implementing a simple round robin architecture in real time operating system by calculating dynamic time slice which depends on three aspects i.e, range, total no of processes in the system, priority of the process & total priority assigned in the system. Yaashuwanth C & R.Ramesh[6] proposed a new method to schedule the processes using scheduling component which is calculated by the help of task period component,task deadline component & priority component. Rami J. Matarneh[4] has proposed an modified round robin where time quantum is adjusted after each iteration by calculating the median of remaining burst time. Abdurazzag Ali Aburas, Vladimir Miho[3] proposed an non-preemtive dynamic priority based scheduling using fuzzy logic.

# III. MATERIALS AND METHODS

#### A. CLUSTERING

The processes are clustered randomly into *k* clusters using "The rule of the Thumb" i.e.  $k \approx \sqrt{n/2}$  with *n* as total number of processes. The *k*- median clustering algorithm can be used to cluster the processes iteratively into dense clusters suitable for scheduling. The median is obtained by arranging the processes in increasing order of their burst time and choosing the burst time of the middle process. If the number of processes is even the middle two processes are chosen and the mean of the burst time of the two processes is taken as the median burst time [4].

$$M = \begin{cases} Y_{(n+1)/2}, \text{ if n is odd} \\ \\ 1/2 [Y_{(n/2)} + Y_{(1+n/2)}], \text{ if n is even} \end{cases}$$

M=median

.

Y=number located in the middle of a group

of numbers arranged in ascending order

n=number of processes

Thus *k*-medians are obtained. The variance of the burst time of each process from the k-medians is calculated and the process is assigned to the cluster with minimum variance.

 $Var(X_i) = E[(X_i - M_{ci})^2]$ 

X=Burst time of the  $i^{th}$  process  $M_{cj}$ =Median of  $j^{th}$  cluster

E represents expected value

This process is carried out iteratively till the medians become constant, obtaining dense clusters fit for scheduling. The weight for each cluster can be found on the basis of the priorities, burst time and waiting time of the processes in a cluster [2]. Initially the WT for all the clusters is 0.

$$\begin{split} & W_{cj} = \frac{WT + \sum P_m}{\sum BT_m} \\ & WT = Waiting time of each cluster \\ & W_{cj} = Weight of the j^{th} cluster \\ & \sum P_m = Sum of User Assigned Priorities of m Processes in j^{th} cluster \\ & \sum BT_m = Sum of burst time of m processes in j^{th} cluster \end{split}$$

#### B. SCHEDULING

The heaviest cluster is chosen to be serviced first. The processes within the cluster are scheduled using Round Robin algorithm based on the Response Ratio [3] and Service Time of each process.

ST=P×BT

$$RR = \frac{WT + ST}{ST}$$

ST=Service time of each process P=User assigned Priority of each process RR=Response Ratio of each process BT=Burst Time of each process

The process with highest Response Ratio is chosen to be executed first in order to prevent starvation due to ageing. The Round Robin algorithm with dynamic time quantum is then followed to schedule the processes within the cluster.

# C. PSEUDOCODE OF PROPOSED ALGORITHM

8) for (i=1 to n){ Calculate variance of BT of ith process from the median of each cluster  $Var(X_{ij}) = E[(X_i - M_{cj})^2]$ 9) if  $Var(X_{ij}) > Var(X_{ij+1})$  $X_{ij}=X_{ij}+1//$  Send  $X_{ij}$  to  $(j+1)^{th}$  cluster// j=j+1// Set cluster number to j+1// 10)GOTO step 2 11)Find the weight of each cluster  $W_{cj} = \frac{WT + \sum P_m}{-}$  $\Sigma BT_m$ 12) While ready queue  $\neq$ NULL 13) Choose the heaviest cluster 14)Choose the process with highest RR within the cluster ST=P×BT WT+ST RR = -ST 15) Use the Round Robin algorithm to schedule the Processes Calculate Dynamic Time Slice for each process  $(\mathbf{R}) \times (\mathbf{N})$ DTS=- $(Pr) \times (P)$ N=Total no of processes in the cluster  $(Max \; BT \; in \; the \; j^{th} \; Cluster + Min \; Burst \\ Time \; the \; j^{th} \; cluster)$ R =2 16) GOTO Step 12 } 17) End

# D. FLOWCHART OF PROPOSED ALGOTITHM



# IV. RESULTS AND DISCUSSIONS

The following experiments were conducted on processes with burst time chosen randomly between 1 to 100 in MATLAB. The unit of burst time is taken to be milliseconds. The pesudo-code is valid for any number of processes, however for simplicity, we have taken 50 processes. The priorities are assigned by the user which is considered external priority of a process. This external priority is particularly useful in an envionment where certain important tasks need to be executed first and thus are assigned higher priorities.

We experimented with 50 processes and using of the "rule of the thumb", the number of clusters is found to be k=5. After applying k-medians clustering the lowest burst time processes form the highest weight cluster and highest burst time processes form the lowest weight cluster according to the weight calculation formula. The highest weight cluster is taken for scheduling first then the next highest cluster. This process continues until all the processes are scheduled with the help of round robin by calculating dynamic time slice.

PROCESSES	BT	PRIORITY
D1	5	1
P2	85	8
P3	27	2
P4	36	9
P5	7	3
P6	9	3
P7	77	2
P8	100	6
P9	3	5
P10	20	7
P11	10	2
P12	61	2
P13	14	6
P14	69	1
P15	24	4
P16	56	8
P17	89	8
P18	4	2
P19	6	5
P20	72	7
P21	1	3
P22	78	2
P23	11	6
P24	34	5
P25	29	7
P26	13	2
P27	66	9
P28	88	9
P29	68	1
P30	24	4
P31	47	7
P32	96	8
P33	67	2
P34	94	9
P35	74	5
P36	38	3
P37	8	1
P38	89	7
P39	22	4
P40	33	3
P41	12	2

Table 1: Processes, their Burst Time and Priority

P42	90	2
P43	78	6
P44	52	1
P45	91	4
P46	20	1
P47	18	8
P48	51	2
P49	46	9
P50	73	3

#### Table 2: Clustering Of Processes

CLUSTERS	MEDIAN	BURST TIME
Cluster 1	4	1,3,4,5,6,7
Cluster 2	78	66,67,68,69,72,73, 74,77,78,78,85,88, 89,89,90,91,94,96,100
Cluster 3	24	18,20,20,22,24,24, 27,29,33,34,36
Cluster 4	51	38,46,47,51,52,56,61
Cluster 5	11	8,9,10,11,12,13,14

# Table 3:Calculation of Dynamic Time Slice

PROCESS	BT	PRIORITY	ST	DTS
P1	5	1	5	4
P2	85	8	680	11
P3	27	2	54	14
P4	36	9	324	3
P5	7	3	21	2
P6	9	3	27	4
P7	77	2	154	42
P8	100	6	600	14
P9	3	5	15	1
P10	20	7	140	4
P11	10	2	20	1
P12	61	2	122	25
P13	14	6	84	2
P14	69	1	69	83
P15	24	4	96	7

P16	56	8	448	7
P17	89	8	712	11
P18	4	2	8	2
P19	6	5	30	1
P20	72	7	504	12
P21	1	3	3	2
P22	78	2	156	42
P23	11	6	66	2
P24	34	5	170	6
P25	29	7	203	4
P26	13	2	26	6
P27	66	9	594	10
P28	88	9	792	10
P29	68	1	68	83
P30	24	4	96	28
P31	47	7	329	8
P32	96	8	768	11
P33	67	2	134	42
P34	94	9	846	10
P35	74	5	370	17
P36	38	3	114	17
P37	8	1	8	11
P38	89	7	623	12
P39	22	4	88	7
P40	33	3	99	9
P41	12	2	24	6
P42	90	2	180	42
P43	78	6	468	14
P44	52	1	52	50
P45	91	4	364	21
P46	20	1	20	27
P47	18	8	144	4
P48	51	2	102	25
P49	46	9	414	9
P50	73	3	219	28



Fig 1: Processes clustered after  $1^{st}$  round of iteration



Fig 2: Processes clustered after 2nd round of iteration



Fig 3: Processes clustered after 3<sup>rd</sup> round of iteration



Fig 4: Processes clustered after 4th and final round of iteration

Table 1 represents the number of processes, burst time of processes and user assigned priority of each process. Table 2 represents the final clusters and medians. Table 3 represents scheduling of processes using Round Robin algorithm with the Dynamic Time Slice, service time and response ratio being the deciding factor for choosing the next process for execution. The time slice is allocated using the formula given in the pseudo-code.

Fig 1 shows the clustering of processes after  $1^{st}$  round of iteration. Fig 2 shows clustering after  $2^{nd}$  round of iteration. Fig 3 shows clustering after  $3^{rd}$  round of iteration and Fig 4 show the final clustering of processes after  $4^{th}$  round of iteration.

# V. EXPERIMENTAL ANALYSIS

$\begin{array}{c c c c c c c c c c c c c c c c c c c $	P1   P5   P21   P18   P9   P19   P1   P5   P18   P9   P19   P5   P9   P	P19 P5 P19 P19
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	22 23 24 25
Fig.5.Gant Chart For Cluster 1     P37   P11   P41   P26   P6   P23   P13   P26   P6     0   8   14   20   26   30   32   34   38   44   50   54   56   58   59   60   62     P23   P13   P23   P13   P23   P13   P13   P13   P13   P36     P40   P39   P15   P30   P24   P10   P25   P47   P4   P3   P40   P39   P15     0   20   34   43   50   57   64   70   74   78   82   85   98   107   114   121     P30   P25   P47   P4   P24   P10   P25   P47   P4   P24   P10   P25   P47   P4   P24   P14   P24   P14   P24   P14   P24   P26   200   200   210 <th< td=""><td>26</td><td></td></th<>	26	
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	Fig 5:Gantt Chart For Cluster 1	
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	P37 P11 P41 P26 P6 P23 P13 P11 P41 P26 P6 P2	3 P13 P26 P6
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	0 8 14 20 26 30 32 34 38 44 50 54 56	58 59 60 62
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$	P23 P13 P23 P13 P23 P13 P23 P13 P13 P13	
Fig 6: Gamt Chart For Cluster 5   P46 P3 P40 P39 P15 P30 P24 P10 P25 P47 P4 P3 P40 P39 P15   0 20 34 43 50 57 64 70 74 78 82 85 98 107 114 121   P30 P24 P10 P25 P47 P4 P40 P39 P15 P30 P24 P10 P25 P47 P4 P24 P4 </td <td>64 66 68 70 72 73 75 77</td> <td></td>	64 66 68 70 72 73 75 77	
P46   P3   P40   P39   P15   P30   P24   P10   P25   P47   P4   P3   P40   P39   P15     0   20   34   43   50   57   64   70   74   78   82   85   98   107   114   121     P30   P24   P10   P25   P47   P4   P40   P39   P15   P30   P24   P10   P25   P47   P4   P24   P24   P24   P24   P24   P24   P24   P24   P24   P25   P4   P24   P44   P4   P2   P24   P20   P23   P20	Fig 6: Gantt Chart For Cluster 5	
0 20 34 43 50 57 64 70 74 78 82 85 98 107 114 121   P30 P24 P10 P25 P47 P4 P40 P39 P15 P30 P24 P10 P25 P47 P4   128 134 138 142 146 149 158 165 172 179 185 189 193 197 200   P40 P39 P15 P30 P24 P10 P25 P47 P4 P24 P20 226 253 259   P25 P4 P24 P25 P4 P25 P4 P4 P4 P4 P4   263 266 270 274 277 278 281 284 287 290 293   F44 P48 P12 P36 P31 P16 P49 P44 P48 P12 P36 P31 P16 P49 P41 P44 P48 P17 P32 P33 <td>P46   P3   P40   P39   P15   P30   P24   P10   P25   P47   P4   P3</td> <td>P40 P39 P15</td>	P46   P3   P40   P39   P15   P30   P24   P10   P25   P47   P4   P3	P40 P39 P15
P30   P24   P10   P25   P47   P4   P40   P39   P15   P30   P24   P10   P25   P47   P4     128   134   138   142   146   149   158   165   172   179   185   189   193   197   200     P40   P39   P15   P30   P24   P10   P25   P47   P4   P24   P10   P25   P47   P4   P24   P26   P25   P47   P4   P24	0 20 34 43 50 57 64 70 74 78 82 85 9	98 107 114 121
128 134 138 142 146 149 158 165 172 179 185 189 193 197 200   P40 P39 P15 P30 P24 P10 P25 P47 P4 P24 P10 P25 P47 P4 P44 P4 <td>P30   P24   P10   P25   P47   P4   P40   P39   P15   P30   P24   P1</td> <td>0 P25 P47 P4</td>	P30   P24   P10   P25   P47   P4   P40   P39   P15   P30   P24   P1	0 P25 P47 P4
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$\begin{array}{c c c c c c c c c c c c c c c c c c c $	P40   P39   P15   P30   P24   P10   P25   P47   P4   P24   P10   P2	25 P47 P4 P24
$\begin{array}{ c c c c c c c c c c c c c c c c c c c$	206 207 210 213 219 223 227 231 234 240 244	248 250 253 259
$\begin{array}{c c c c c c c c c c c c c c c c c c c $	P25   P4   P24   P25   P4   P25   P4   P4   P4	P4 P4
Fig 7: Gantt Chart For Cluster 3   P44 P48 P12 P36 P31 P16 P49 P44 P48 P12 P36 P31 P16 P49 P44   0 50 75 100 117 125 132 141 143 168 193 210 218 225 234 235   P12 P36 P31 P16 P49 P31 P36 P31 P32 P33 P34 P33 P37 P32 P35 P35 P35 P35 P35 <t< td=""><td>263 266 270 274 277 278 281 284</td><td>287 290 293</td></t<>	263 266 270 274 277 278 281 284	287 290 293
P44   P48   P12   P36   P31   P16   P49   P44   P48   P12   P36   P31   P16   P49   P48     0 50   75   100   117   125   132   141   143   168   193   210   218   225   234   235     P12   P36   P31   P16   P49   P31   P16   P30   P31   P31   P31   P31   P31   P31	Fig 7: Gantt Chart For Cluster 3	
114 112 112 112 113 114 144 144 168 193 210 218 225 234 235   P12 P36 P31 P16 P49 P31 P	P44 P48 P12 P36 P31 P16 P49 P44 P48 P12 P36 P3	R1 P16 P49 P48
P12 P36 P31 P16 P49 P31 P16 P49 P31 P16 P49 P31 P16 P49 P16   246 250 258 265 274 282 289 298 306 313 322 329 336 337 344   P16 P16 P16 P35 P43 P8 P20 P38 P2 P17 P32   0 68 137 179 221 263 305 333 354 371 385 399 411 423 434 445 456   P27 P28 P34 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32   466 476 486 511 546 582 624 652 673 690 704 718 730 742 753 764   P32 P27 P28 P34 P42 P50 P45 P35 P43 P8	050 75 100 117 125 132 141 143 168 193 210	218 225 234 235
P12 P30 P31 P10 P49 P31 P10 P	D12 D26 D21 D16 D40 D21 D16 D40 D21 D16 D40 D21	21 D16 D40 D16
P16 P16   351 358   Fig 8: Gantt Chart For Cluster 4   P29 P14 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32   0 68 137 179 221 263 305 333 354 371 385 399 411 423 434 445 456   P27 P28 P34 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17   466 476 486 511 546 582 624 652 673 690 704 718 730 742 753 764   P32 P27 P28 P34 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27   775 785 795 805 811 828 849 866 870 884 896	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	329 336 337 344
Fig 8: Gantt Chart For Cluster 4   P29 P14 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32   0 68 137 179 221 263 305 333 354 371 385 399 411 423 434 445 456   P27 P28 P34 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32   466 476 486 511 546 582 624 652 673 690 704 718 730 742 753 764   P32 P27 P28 P34 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27 P32 P37 P43 P8 P20 P38 P2 P17 P32 P27 P38 P3 P17 P32 P27 P28 P34	D16 D16	
Fig 8: Gantt Chart For Cluster 4   P29 P14 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32   0 68 137 179 221 263 305 333 354 371 385 399 411 423 434 445 456   P27 P28 P34 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17   466 476 486 511 546 582 624 652 673 690 704 718 730 742 753 764   P32 P27 P28 P34 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27 P37 765 795 805 811 828 849 866 870 884 896 908 919 930 941 951   P28	351 358	
P29 P14 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32   0 68 137 179 221 263 305 333 354 371 385 399 411 423 434 445 456   P27 P28 P34 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17   466 476 486 511 546 582 624 652 673 690 704 718 730 742 753 764   P32 P27 P28 P34 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27   75 785 795 805 811 828 849 866 870 884 896 908 919 930 941 951   P28 P34 P45 P35	Fig 8: Gantt Chart For Cluster 4	
0 68 137 179 221 263 305 333 354 371 385 399 411 423 434 445 456   P27 P28 P34 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17   466 476 486 511 546 582 624 652 673 690 704 718 730 742 753 764   P32 P27 P28 P34 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27   775 785 795 805 811 828 849 866 870 884 896 908 919 930 941 951   P28 P34 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27 P28 P34 P45 P35 961 971 992 1009 1023	P29   P14   P33   P7   P22   P42   P50   P45   P35   P43   P8   P20   P35	88 P2 P17 P32
P27 P28 P34 P33 P7 P22 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17   466 476 486 511 546 582 624 652 673 690 704 718 730 742 753 764   P32 P27 P28 P34 P42 P50 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27   775 785 795 805 811 828 849 866 870 884 896 908 919 930 941 951   P28 P34 P45 P35 P43 P8 P20 P38 P2 P17 P32 P27 P28 P34 P45 P35 943 P45 P35 943 P45 P35 943 P45 P35 943 P45 P35 944 P45 P35 944 P45 P35 944 P45 P35 944 P45	0 68 137 179 221 263 305 333 354 371 385 399 411	423 434 445 456
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Fig 9:Gantt Chart For Cluster 2

Table 4: Calculation of TAT,WT & CS of each cluster Comparision of AvgTAT, AvgWT and CS of 5 clusters

	Avg TAT	Avg WT	CS
Cluster 1	17	12.16	17
Cluster 2	1060.7	980.05	103
Cluster 3	208.09	181.18	55
Cluster 4	271	211.7	31
Cluster 5	51.2	40.2	23





The experimental result shows that clustering of similar processes into dense clusters providing a better and easy scheduling of processes taking into account external priority and using Dynamic Round Robin algorithm which is suitable for real time systems. Clustering strategy will boost performance far beyond any tuning that might be done after the processes are loaded and executed will reduce your working set size by keeping similar size job into one cluster; yield higher data density; and reduce the overhead and save memory space [6].Heaviest weight

Cluster has lowest turn around time and lowest waiting time in comparison to other clusters. The above method can also be implemented in multi-processor system where the heavier clusters can be executed by high speed processors, thus increasing efficiency and decreasing the time complexity of today's complex working of operating system.

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