

Workflow Scheduling Under Secure Cloud Environment using MPSO-SA



Babita Bhagat, P.Sanyasi Naidu

Abstract— *In the distributed data-intensive computing environment, relegating certain assignments to specific machines in a protected way is a major test for the employment planning issue. The unpredictability of this issue increments with the size of the activity and it is hard to understand viably. A few metaheuristic calculations including particle swarm optimization (PSO) strategy and variable neighborhood particle swarm optimization (VNPSO) system are utilized to tackle the employment planning issue in distributed computing. While allocating assignments to the machines, to fulfill the security requirements and to limit the cost capacity, we proposed an altered PSO with a scout adjustment (MPSO-SA) calculation which utilized a cyclic term called change administrator to get the best cost capacity. The exhibition of the proposed MPSO-SA booking component is contrasted and the Genetic calculation (GA), PSO and VNPSO systems and the exploratory outcome demonstrate that the proposed technique diminishes the likelihood of hazard with security requirements and it has preferable intermingling property over the current conventions.*

Keywords—Cloud computing, metaheuristic, GA, PSO and VNPSO.

I. INTRODUCTION

Distributed computing is an enormous scale dispersed figuring persuaded to satisfy the necessities of modest and costly framework tasks, in which the foundations are realistic in a compensation for each utilization framework and can supply dynamic scaling as for the prerequisites of work process applications. On account of the multifaceted nature of logical registering frameworks, such work process applications have turned out to be enormous information applications, which lessen the intricacy of huge-scale information by decaying the application to littler undertakings and that can be handled in a grouping request by methods for booking. Planning enables the assets to ideally assign among the given undertakings in a fixed time to accomplish the ideal nature of administration. Officially, the booking issue includes errands that must be planned on assets subject to certain imperatives to streamline some target work. By and large, the issue of mapping undertakings on evidently unbounded figuring assets in distributed computing has a place with a classification of issues known as NP-difficult issues.

II. LITERATURE REVIEW

The principle issue of cloud administrations is security which makes the association waver most with regards to moving business remaining tasks at hand from private Cloud into open Cloud. A few looks into have made for proficient assignment planning for the cloud, however undertaking booking issues are still NP-complete. The majority of the errand booking calculations utilized in distributed computing are standard based in light of the fact that they are anything but difficult to execute. Standard-based calculations perform inadequately with regards to complex undertaking planning issues. The most well-known metaheuristic strategies connected to assignment booking issues in the lattice and distributed computing are GA, PSO, and Ant Colony Optimization (ACO). PSO joins quicker and gets preferred arrangements over GA and ACO due to its exploratory capacity for finding ideal arrangements. Attributable to the better execution of PSO over ACO and GA, an altered variant of PSO with scout adjustment has been utilized for benchmarking the proposed calculation.

In 2015, Lingfang Zeng tended to the issue of security in work process booking and present a Security-Aware and Budget-Aware work process planning technique (SABA) to acquire shorter makespan and verified planning under some spending imperative. They additionally planned a proficient technique to bunch errand dependent on information reliance that diminishes the expense and time of recovering information in the cloud framework. The trial results demonstrated that SABA is viable and productive in creating cost and execution of work process booking; additionally, gives better security in cloud condition.

In 2016, Zhongjin proposed security and cost mindful booking (SCAS) calculation for logical work processes which considers the heterogeneous undertaking with various qualities like as information concentrated, memory-serious or calculation escalated. SCAS calculation is, for the most part, used to limit the expense of execution at which it faces the due date and hazard rate requirements. They utilized the coding approach of PSO to get the answer for a multi-imperative and multi-dimensional streamlining issue in work process planning.

In 2016, Saima Gulzar Ahmad has changed a hereditary assessment based procedure to another crossover hereditary calculation (HGA) that gives an ideal arrangement in less measure of time. This calculation uses the assets in enormous sum by upgrade the heap adjusting at the season of execution. At last, the presentation of HGA calculation is investigated utilizing orchestrated datasets and genuine application work processes.

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In 2015, Mala Kalra and Sarbjeet Singh have introduced a summed up methodology for booking in cloud computing. A nitty-gritty study and similar investigation of different planning calculations mostly the three prevalent metaheuristic calculations: Ant Colony Optimization (ACO), GA and PSO, and two novel strategies: League Championship Algorithm (LCA) and BAT calculation are talked about in this paper.

In 2016, Mohammed Abdullahi has discovered an ideal answer for proficient errand planning utilizing Symbiotic Organism Search (SOS) which is a metaheuristic improvement procedure. The SOS calculation for undertaking planning is executed utilizing makespan, reaction and level of lopsidedness among different virtual machines. At last, the measurable reports of the outcomes are contrasted and SAPSO utilizing some hugeness test.

III. WORK FLOW APPLICATION MODEL FOR CLOUD SYSTEM

A. Work flow Architecture

Cloud condition comprises an accumulation of virtual machines that give administrations, for example, Infrastructure administration, Platform administration, and Software administration. Besides, it has some significant highlights like virtualization, appropriation, and flexibility. As a rule, distributed computing maps the applications to the cloud virtual machines utilizing virtualization technique and so booking of undertakings for each mapped machine is completed in the application layer and the virtual layer. The planning alludes to a lot of principles to arrange the attempts to be performed by the organized systems. Here, the complexities of enormous scale issues are decreased by work process application and consequently work process booking discovers prime significance. The primary point of the work process booking is to limit the makespan by allocating certain assignments to specific machines in an appropriate manner. Work process engineering which shows in Fig. 1 comprises a bundle of components that must be a plan for a consecutive request.

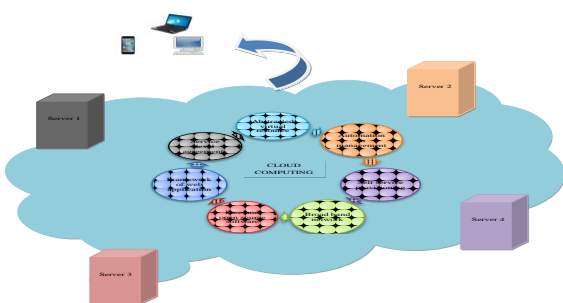


Fig. 1. Workflow architecture of cloud computing environment

B. System Model

The PC application which procedures an enormous volume of information is named as information escalated that gives a large portion of its handling time to I/O and after that, it controls the information. In information escalated application, the activity and the assets must meet the necessities of the procedure stream, information move and information get to, cost of fulfillment, security limitations, adaptability, and accessibility. The information escalated

application is improved by the work process, that breakdown the application into a littler assignment and is handled in an arrangement request to accomplish the outcome. In the information concentrated figuring condition, the information asset (database is associated with the server) is associated with the computational unit (Super-PC or some other machines) by an interfacing join which speaks to the relating transmission capacity and the limit of the registering unit relies on the quantity of focal handling unit (CPU), memory and extra room and other specialization. The preparing velocity of each machine is communicated as the number of cycles per unit time.

Give us a chance to consider the information concentrated application in cloud condition with five processing units and five information assets and every one of the segments must interface straightforwardly or by implication with one another as appeared in Fig. 2. However, the segments of information escalated figuring can be summed up as processing units in which each registering unit furnishes the registering administration with the security rank and information assets in which every datum asset gives the information administration the security rank and both the security positions are ordinarily named as.

A lot of errand/activity establishes a vocation that must be performed on a machine with pivotal preparing limitations. Three of such significant handling requirements are as per the following,

1. Work process limitations: Arranging a lot of errand or activity for a particular application has alluded as a work process and each undertaking can be performed after the fulfillment of the past assignment in succession.
2. Preparing length: The quantity of cycles that require for the fruition of a task is named as handling length. An undertaking must perform inside this handling length.
3. Security requirements: If a work process application comprises of occupations and the activity comprises of a lot of assignment/task. To diminish the trouble, every one of the activities is consolidated together as, where alludes to the absolute number of undertakings and it is resolved to utilize eq. (1),

$$N_T = \sum_{q=1}^K |T_q| \quad (1)$$

In the eq. (1), the cardinality term $|T_q|$ refers to a set of tasks in q^{th} job.

The security demand of the computing service SD_{cs,T_i} and the security demand of the data service SD_{ds,T_i} are commonly termed as S_T .

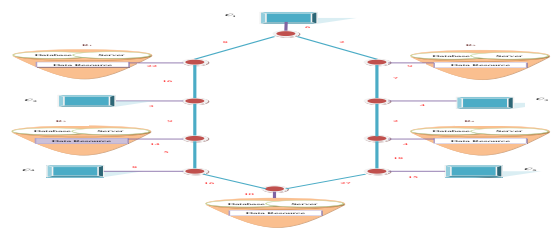


Fig. 2. Data-intensive computing environment with five computing units and five data resources .

C. Security Constraints

In information concentrated registering condition, three security modes are accessible for occupation planning process: Secure mode: Generally, the tasks that fulfill the security prerequisites and that are available on the processing units are planned for this protected mode. This mode is a customary methodology for planning work. In this mode, if the security request of the undertaking is lesser than or equivalent to the security rank of the processing unit or information asset for example at that point just the activity is distributed to the relating figuring unit or information asset. Secure mode: All the possible dangers are evacuated by booking the undertaking just to the accessible registering unit or accessible information asset. This mode is an aggressive and intense methodology for planning the activity.

γ -risky mode: In this mode, the task is scheduled to available data resources or machinestaking at most γ risk, in which γ represents the measurement of probability with $\gamma=0$ and $\gamma=1$ for the corresponding secure and risky modes.

Generally, the secure mode is very expensive to attain, so risk mode and γ -risky mode is commonly employed to solve job scheduling problem. In the cloud computing environment, the security levels are approached by a qualitative or fuzzy scale which consists of five levels namely, very high, high, medium, low and very low.

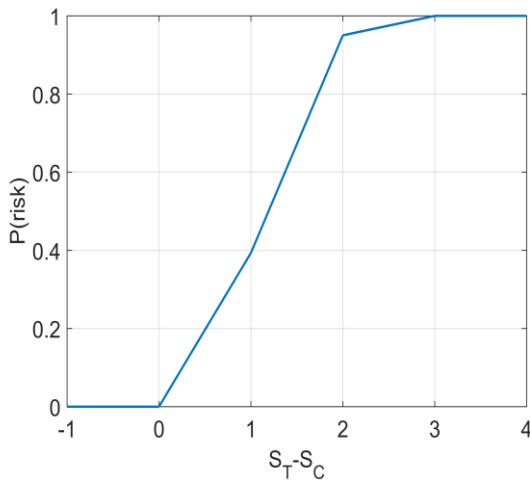


Fig. 3. Graphical representation of risk probability

The probability of risk for the security constraint model is defined in eq. (2) and it is graphically represented in Fig. 3

$$Pr ob_{risk} = \left\{ \begin{array}{ll} 0 & \text{if } S_T - S_C \leq 0 \\ 1 - e^{-0.5(S_T - S_C)} & \text{if } 0 < S_T - S_C \leq 1 \\ 1 - e^{-1.5(S_T - S_C)} & \text{if } 1 < S_T - S_C \leq 2 \\ 1 & \text{if } 2 < S_T - S_C \leq 5 \end{array} \right\} \quad (2)$$

A figuring unit with the condition, or is considered as a protected machine and here, the likelihood of hazard is zero. On the off chance that a planned activity is relegated to a machine with the condition, at that point the likelihood of hazard must be lesser than half and in the event that an assignment is given to a registering unit with, at that point the booked errand will be done by the machine. The booked

activity is executed later, if, yet it will be executed before as far as possible. At the point when the assignment can't be practiced and it must be rescheduled. In this condition, planning has turned into a significant worry to perform in the information escalated cloud condition.

IV. SECURE WORK FLOW SCHEDULING

A. Scheduling Model

Let us consider a work flow application that is executed by the machine with the processing speed $\{P_{S_1}, P_{S_2}, \dots, P_{S_{N_c}}\}$ and the processing length $\{l_1, l_2, \dots, l_{N_r}\}$ and finally it is subjected to a set of security constraint, $CS = \{S_T, S_C\}$.

Fig. 4 shows the Directed Acyclic Graph (DAG) based workflow representation for job scheduling. The workflows are usually designed as Directed Acyclic Graphs(DAGs), $G\{T, E\}$ in which vertices, $T = \{T_1, T_2, \dots, T_{N_r}\}$ denotes the individual task of the workflow and edges E of the graph denotes the precedence or data dependencies among the task.

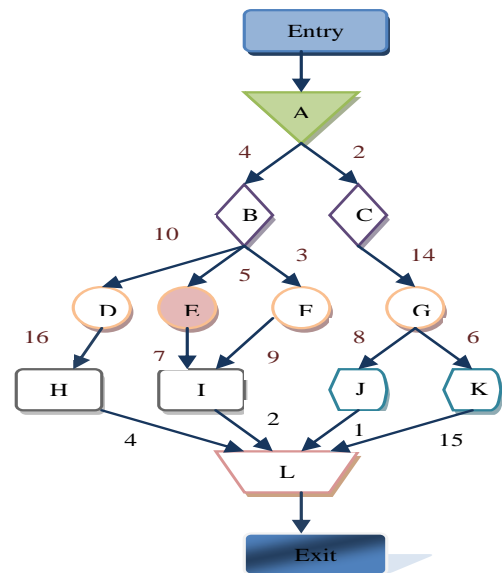


Fig. 4. DAG based workflow representation for job scheduling.

Definition 1: An edge of the DAG is spoken to as, which infers that the errand can't begin to process until the undertaking finishes its procedure and send the outcome to. Allow us to consider, is a lot of forerunner of the activity and is a lot of successor of the assignment. Some supposition that is made for employment planning for processing condition, as it is an NP-difficult issue,

- (a) A successor errand begins to perform following its antecedent finishes its procedure, as it demonstrates the accessibility of the machine.
- (b) Only one undertaking/task can be taken care of by a machine at any given moment.
- (c) Each assignment must be taken care of by any of the machines.
- (d) The undertaking can't be reprocessed, when it finishes its procedure effectively.



(e) The span for arrangement and move are uniform.

(f) The activities are free.

From Fig. 4, the relations between the assignments are connected by a stream framework. In the event that the edge exists in the chart, at that point gather its weight, on the off chance that it doesn't exist, at that point is set to.

The information asset conditions of assignments are found by the recovery lattice, in which the component shows the recovery time for the errand which is preparing the information recovery from the information asset. Additionally, the throughput rates are determined by the two frameworks and, in which the component speaks to the limit of the association interface between the processing units or machines and, and the limit of the association connect between the registering unit and information asset is spoken to by. The procedure consummation time for every task is assessed by the total of the span for the three exercises: time for recover information, the time is taken to include information and the expected time to execute the activity on the accessible machine.

Let us consider a feasible solution, $Y = \{Y_1, Y_2, \dots, Y_{N_t}\}$, in which Y_i indicate the serial number of the computing unit to which the task T_i is allocate, then the completion time CT_{T_i} of the task T_i on the machine C_{Y_i} is evaluated using the eq. (3)

$$CT_{T_i} = \sum_{x=1}^{N_r} f_{x,i} u_{Y_i, Y_i} + \sum_{z=1}^{N_i} r_{i,z} v_{Y_i, z} + l_i / P_{S_{Y_i}} \quad (3)$$

The maximum completion time (makespan) CT_{max} is computed using the eq. (4),

$$CT_{max} = \max \{ \sum CT_{C_i} \} \quad (4)$$

The flowtime i.e, the sum of the completion time of the solution is evaluated using the eq. (5),

$$CT_{sum} = \sum_{i=1}^c (\sum CT_{C_i}) \quad (5)$$

The above eq. (4)&eq. (5) are used as a performance criterion for job scheduling problem in computing environment. If CT_{sum} gets minimize, it tends to reduce the execution time, and so the largest task requires a long time to process, whereas minimizing CT_{max} seeks to increase the execution time, and hence no task requires too long time to execute. As a result, to obtain an equal balance between these two criteria, a weighted aggregation is employed as in eq. (6),

$$f = W_1 \{CT_{max}\} + W_2 \{CT_{sum}\} \quad (6)$$

In the eq. (6), W_1 and W_2 are non-negative weights, this can be fixed or adjusted but the sum of both weights should be unity.

B. Conventional Heuristics

1. Canonical PSO model:The standard PSO calculation was enlivened from the conduct of winged animal running and fish tutoring. The two primary factors that describe the status of a molecule in the pursuit space are its position and speed. Eq. (7) is utilized to refresh the speed and Eq. (8) is utilized to decide the new position by including the past position and the new speed.

$$h_{i,q}(t+1) = lh_{i,q}(t) + ac_1 r_1 (g_{i,q}^*(t) - g_{i,q}(t)) + ac_2 r_2 (g_q^*(t) - g_{i,q}(t)) \quad (7)$$

$$g_{i,q}(t+1) = g_{i,q}(t) + h_{i,q}(t+1) \quad (8)$$

In eq. (7) and eq. (8), and are the speed of the molecule, the position of the molecule, past best position and the best area among all particles in the populace respectively. Here and are the arbitrary capacities inside a range and are the two positive consistent parameters called increasing speed coefficients, which lessen the greatest advance size of the molecule. The alluring outcomes can be acquired by choosing with and mean the latency factor while advancing the calculation it's worth gets diminishes from 1.0 to 0.

So as to screen the molecule in hunt space zone, the most extreme development of speed in one cycle and the situation of the molecule is restricted inside the range as in eq. (9) and eq. (10) as,

$$h_{i,q} = \text{sign}(h_{i,q}) \min(|h_{i,q}|, h_{max}) \quad (9)$$

$$g_{i,q} = \text{sign}(g_{i,q}) \min(|g_{i,q}|, g_{max}) \quad (10)$$

Here, h_{max} is the maximum velocity and g_{max} is the maximum solution variable.

Generally, the PSO algorithm has two basic updating models namely *gbest* model and *pbest* model. According to eq. (7), the global best (*gbest*) position $g_q^*(t)$ is the best particle (obtained by means of fitness function) till the t^{th} iteration and the local best solution (*pbest*) $g_{i,q}^*(t)$ is the best i^{th} particle in the t^{th} iteration and it is given in eq. (11),

$$g_i^* = \arg \min_{i=1}^N (f(g_i^*(t-1)), f(g_i(t))) \quad (11)$$

The *gbest* and *pbest* alongside and control the impact of new molecule speed. By and large, *gbest* model combines rapidly than the *pbest* model, and furthermore, it tends to limit in neighborhood optima yet *pbest* model tends to stream around the nearby optima.

2. Variable neighbourhood particle swarm optimization algorithm (VNPSO):

In eq. (7) and eq. (8), and are the speed of the molecule, the position of the molecule, past best position and the best area among all particles in the populace respectively. Here and are the arbitrary capacities inside a range and are the two positive consistent parameters called increasing speed coefficients, which lessen the greatest advance size of the molecule. The alluring outcomes can be acquired by choosing with and mean the latency factor while advancing the calculation it's worth gets diminishes from 1.0 to 0.

So as to screen the molecule in hunt space zone, the most extreme development of speed in one cycle and the situation of the molecule is restricted inside the range as in eq. (9) and eq. (10) as,

$$v_{e_{i,q}}(t) = Wv_{e_{i,q}} + ac_1 r_1 (g_{i,q}^*(t-1) - g_{i,q}(t-1)) + ac_2 r_2 (g_q^*(t-1) - g_{i,q}(t-1)) \quad (12)$$

$$v_{e_{i,q}} = \begin{cases} v_{e_{i,q}} & \text{if } |v_{e_{i,q}}| \geq v_{e_i} \\ u_d v_{e_{max}} / \rho & \text{if } |v_{e_{i,q}}| < v_{e_i} \end{cases} \quad (13)$$

In eq. (13), speaks to uniform irregular circulation work, is the most extreme speed and is the speed of a molecule with a measurement. The estimation of two parameters and are related to the presentation of the calculation. In the event that is huge, at that point, the time of swaying gets decreased and it expands the likelihood that the particles hop over the neighborhood minima for a similar number of cycles and furthermore it keeps the molecule from merging by keeping the particles in the quick flying state. These procedures are reshaped until it fulfills the end criteria.

C. Proposed MPSO-SA

The proposed MPSO-SA contrasts from PSO and VNPSO by showing the scout adjustment process after the new arrangement focuses are resolved. Give us a chance to start the arbitrary speed and irregular position for every molecule with the measurement. At that point, the wellness capacity of the position can be resolved for every molecule. On the off chance that the position wellness worth is lesser than the wellness capacity of the neighborhood best for every molecule then the position is the nearby best position. At times, the wellness capacity of the position winds up more noteworthy, than the neighborhood best wellness esteem then another strategy is proposed to compute the number of positions to be changed utilizing eq. (14),

$$N_{position} = \max \left\{ \left\lfloor \frac{cycle}{N_{cycle}} * q \right\rfloor, 1 \right\} \quad (14)$$

In the eq. (14) is the round off chairman that gives the estimation of the closest entire number by leaving the sign and fragmentary part, where the amount of cycles is portrayed in set number. From the start, it is set to zero and expanded by one to obtain the best position. Contingent on the number of positions, the change method occurs between the got position and the sporadic position and from that point onward, the new position gets refreshed. At that point, the technique discussed in the authorized PSO model has jumped out to get the best position, which changes in finding the overall best position as given in eq. (15),

$$gbest = \min \{ f(pbest_i) \} = pbest_i^* \forall i \quad (15)$$

The pseudocode of the proposed MPSO-SA algorithm is illustrated in Algorithm 3 as given below:

ALGORITHM 3: PSEUDOCODE OF PROPOSED MPSO-SA ALGORITHM

```

Input : N, I, ac1 and ac2, r1 and r2, h1, gi and cycle = 0
Output: gbest
Begin
  Initialize N number of particles
  For each particle i ∈ N and dimension D,
    Initialize velocity hi and position gi randomly
    Evaluate gi by determining f(gi)
    if f(gi) < f(pbesti) for each particle i, then
      pbesti = gi, for each particle i
      reset cycle = 0
    else
      cycle = cycle + 1
    Find the number of position N_position using eq.(14) and
    update the new position gi.
  end if
  If min { f(pbesti) } ∀ i < f(gbest) then
  
```

```

      gbest = pbesti*
    end if
    Update the particle's position and velocity using eq. (7) and eq.
    (8)
  end for
  t = t + 1
  Until t > Max_iterations
  
```

V. SIMULATION RESULTS

A. Simulation Setup

The reproduction for the employment planning process in distributed computing condition and the execution of proposed MPSO-SA calculation are done in MATLAB R2015a. The outcomes are contrasted and different calculations (GA, PSO, and VNPSO) to quantify the adequacy of the proposed MPSO-SA model regarding assembly, cost statics, and hazard likelihood. The booking issues have been reenacted in three cases by shifting a number of employments and figuring machines. Give the issue of booking 20 employments in 5 figuring units a chance to be case 1 and planning 50 occupations in 10 processing units be case 2 and booking 100 employments in 20 registering units are case 3. The experiment has been kept running in three cases for multiple times more than 100 cycles with various irregular contributions for every calculation. The reproduction is executed by fixing the qualities as,

- a) The size of the particle is twice the number of task/operation.
- b) Inertia weight I is set to 0.9
- c) The acceleration coefficients, ac1 = ac2 = 1.49
- d) Set cycle = 0

B. Statistical Analysis

In this segment, the measurable data of the current and the proposed framework is investigated and the outcomes are arranged in Table I, Table II and Table III. The factual examination of the considerable number of calculations as far as cost capacity has announced that the mean estimation of MPSO-SA of case 1, case 2 and case 3 are 40.74%, 12.45% and 34.01% superior to the traditional PSO model individually.

TABLE I. COST STATISTICAL ANALYSIS FOR CASE 1

	GA	VNPSO	PSO	Proposed MPSO-SA
Best	308.82	67	91.37	10.23
Worst	771.88	312.81	234.9	115.68
Mean	442.35	170.06	132.69	78.631
Median	368.78	192.3	110.56	90.576
Standard deviation	187.39	102.21	59.043	41.429

TABLE II. COST STATISTICAL ANALYSIS FOR CASE 2

	GA	VNPSO	PSO	Proposed MPSO-SA



Best	1206.6	858.53	1022.8	939.16
Worst	1563.2	1270.7	1322.3	1103.1
Mean	1394.1	1084.6	1163.8	1018.8
Median	1389.7	1197	1145.4	1013.2
Standard deviation	134.09	193.65	118.3	58.319

TABLE III. COST STATISTICAL ANALYSIS FOR CASE 3

	GA	VNPSO	PSO	Proposed MPSO-SA
Best	587.19	458.61	618.22	420
Worst	919.68	681.49	846.92	557.68
Mean	804.19	545.7	710.17	468.59
Median	841.86	530.22	682.8	462.2
Standard deviation	126.9	89.495	85.114	53.103

The below graph shows the cost analysis of algorithm when comparing MPSO-SA with other algorithm like GA, PSO, VNPSO. In fig.5 we can identify that cost analysis of MPSO-SA is lesser as compare with any algorithms.

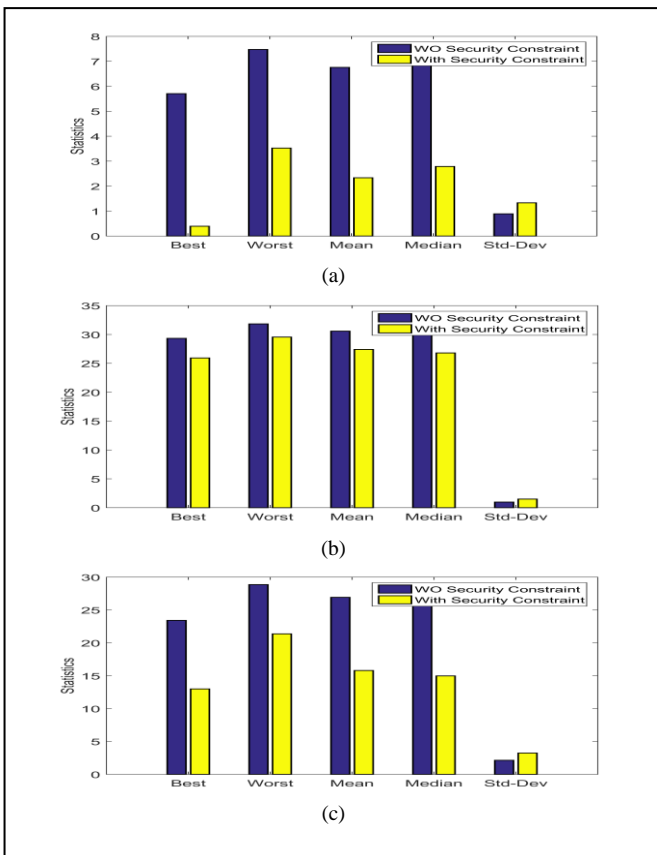


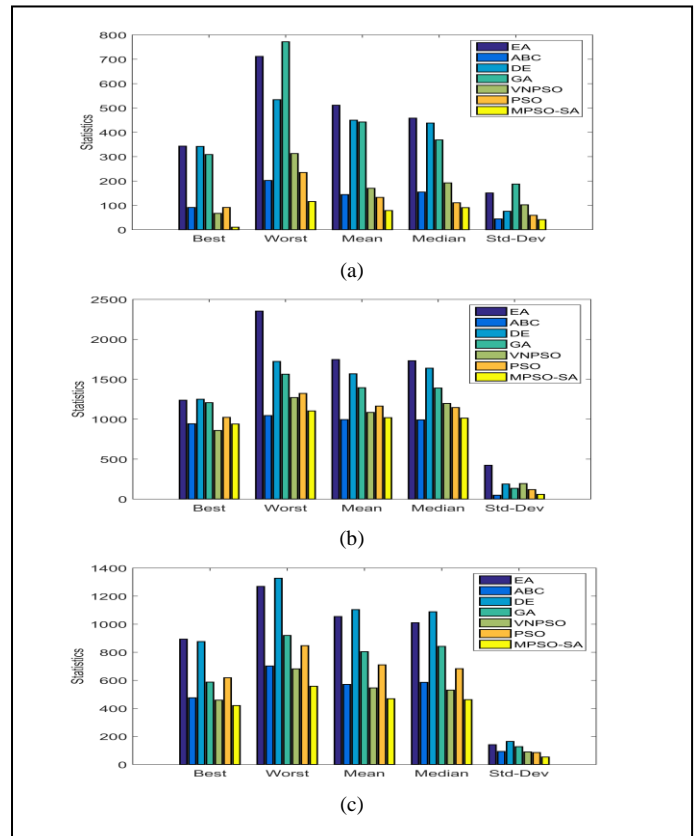
Fig.5 Cost Analysis by considering case1,case2 and case3

The likelihood of hazard for the security requirement model which is clarified in Eqn. (2) is tentatively actualized and delineated in Table IV. The best booking arrangement is gotten by considering the security imperatives that will expand the likelihood of executing the activity with no hazard. Nonetheless, it is seen that the issue that has been illuminated by the proposed MPSO-SA calculation with security limitations is better when contrasting and the arrangement that gotten by the calculation without security requirements in case of mean, middle and standard deviation. In this manner, the security requirement assumes a significant job to get the best answer for the booking issue.

TABLE IV. RISK PROBABILITY STATISTICAL ANALYSIS FORCASE 1, CASE 2 AND CASE 3

	Case 1		Case 2		Case 3	
	without security constraints	with security constraints	without security constraints	with security constraints	without security constraints	with security constraints
Best	5.7045	0.39347	29.323	25.93	23.421	12.997
Worst	7.4743	3.5241	31.83	29.55	28.861	21.373
Mean	6.7564	2.3344	30.577	27.411	26.931	15.8
Median	7.311	2.7869	30.337	26.799	27.861	14.98
Standard deviation	0.8896	1.3311	0.97634	1.496	2.1423	3.2469

The best solutions of Risk Analysis that are obtained by GA, PSO, VNPSO and proposed MPSO-SA for solving the all the cases are discussed in Fig. 6



From the above analysis, it is clear that the best solution for the job scheduling problem is obtained by the proposed MPSO-SA algorithm than the existing algorithms.

VI. CONCLUSION

In this paper, a security imperative model is detailed with occupation booking issue and its exhibition is checked and contrasted and hereditary calculation, PSO and VNPSO calculation.

The exhibition of the proposed MPSO-SA is investigated with the three fundamental properties of planning component, union, cost statics and hazard likelihood statics. The trial consequences of the proposed MPSO-SA is contrasted and GA, PSO and VNPSO and the outcome demonstrates that it gives better cost capacity and generally safe likelihood and the mean estimation of every case are 40.74%, 12.45% and 34.01% superior to the current systems.

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