

A Comparative study on TSP

February 28, 2018

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1 Abstract

Traveling salesmen problem is the most important and familiar problem in order to get the optimal and shortest path to reach at the goal point. To achieve this objective multiple approaches has been developed. In this paper we are comparing different approaches from the collection of many algorithm and approach to present that all of these are providing best results of optimization. But these approaches having best and optimal results.

2 Introduction

Traveling salesman problem(TSP) is the most famous, important and abstracted problem to find the shortest path from a set of nodes locations. This problem is consider as the foundation problem in different fields not either in computer science. This problem focus on route finding algorithm and optimal path planning for vehicle or salesmen with the minimum traveling cost. For this purpose of robots route finding most recently new research approach Stochastic TSP and Orienteering for dynamic system[1]. This approach work on randomly distributed set of locations. In which approach deals with the impact of length on the tours for optimization selection of shortest tour. Because generally in optimal solution finding needs to the computational time given $O(n!)$ where n number of cities. In TSP with the set of few number of location cities determining through linear programming. But in case of complex data set of locations computation can be extremely time intensive

. For such purpose finding new framework also used in MTSP problem with maximum entropy principal base approach[2]. This approach find the shortest route of a salesman from set of routes. In which consider the many constraints also like total distance covered is minimized. For single salesman use heuristic approach with always their starting and ending points are same. This concept presents to a low efficient resources algorithm like deterministic annealing, combinational and resource allocation problem. Selective TSP deals with the single vehicle or sometimes many of them salesman involved. But when more then one vehicle involved then its called routing problem with the profit and minimum travel cost. To solving of this problem purposes a new algorithmic approach Genetic algorithm developed[3]. This consider orienteering problem which used in practical sences. This presented mathematical model for the some subsequent details. Which uses in determining convergence to optimal solution. Genetic algorithm is basically a search technique for computing of exact or approximated optimal solution. This transforms population in natural way of occurring genetic operations such as crossover and mutation. Examples of NP-hard are Classical Search Maps, Simulated Annealing and Artificial neural network. Now GELS GA algorithm approach is design to solving the mTSP with the sufficient and realistic results. TSP in which searching of possible optimized combinations. From complex set of nodes at computational step leads towards more expensive and unrealistic results. So far that purpose of work to find shortest approaches for other GA and ACO heuristic approaches. Hybrid

Ant Colony Optimization approach design to solve the difficult problems. And underperformance of the capability for finding optimal solution through Euclidean Equations[6]. With the help of this equation by change of the root equation through ACO algorithm. It can include the new results as compared to the other equations results which has basic goal of developing.

3 THEROEMS REPRESENTATION IN FIGURE

Theorem 1 (Bounds on TSP_Π). Let Π be a set of constraints, and f be a probability distribution over X , satisfying the assumptions given in Section II-C. Then, if $X_1, X_2, \dots, X_n \in X$ are identically and independently distributed according to f ,

$$TSP_{\Pi}(X_1, X_2, \dots, X_n) = \Theta(n^{(7-1)/7})$$

with very high probability.

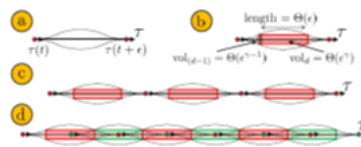
Theorem 2 (Bounds on ORNT_Π). Let Π be a set of constraints, and f be a probability distribution over X , satisfying the assumptions given in Section II-C, and let $\lambda > 0$. Then, if $X_1, X_2, \dots, X_n \in X$ are identically and independently distributed according to f ,

$$ORNT_{\Pi}(X_1, X_2, \dots, X_n; \lambda) = \Theta(\lambda n^{1/7})$$

with very high probability.

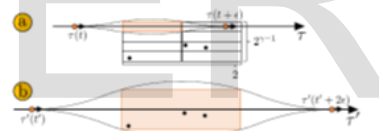
To solve the length of set of n distributed number of cities through which can be obtain the best result of Orienteering with the high probability. These algorithm are simple and provide the approximate optimal solutions for them. In the first theorem special parameter is used to define the length of tour. But in the second theorem is represent the Orienteering of TSP. With the use of these algorithms having the optimal results got.

4 Problem representation of TSP upper Bound

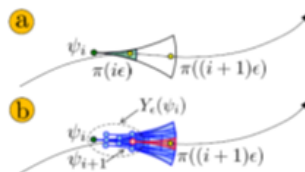


This figure represents the solution of optimal shortest length of tour by using the first theorem which automatically proves the Orienteering and collects all the distributed n number of cities for sub length of tour. Which explain clearly in below figure.

5 SOLUTIONS OF THEOREM



figurer:2 represents the cell merging process for different number of cities path finding



figurer:3 illustrates the initial point to move towards the goal point achieving Multiple Traveling Salesmen and Related Problems A Maximum-Entropy Principle based Approach[2]. This research paper based on two algorithm for solving the multiple traveling salesman problem One

algorithm is MEP and second algorithm Close enough multiple traveling salesman problem. These are basically heuristic approaches with the help of them optimal solution for TSP can be found easily at run time. The CEMTSP equation is given that

6 CEMTSP EQUATION Figure

$$F = -\frac{1}{\beta} \sum_{i=1}^n \log \left(\sum_{j=1}^n e^{-\beta d(x_i, y_j, \rho_i)} \right) + \theta \sum_{j=1}^n d(y_j, y_{j+1})$$

Taking derivative of the free-energy term and setting it to 0, we obtain the update equation for each facility given by

$$y_j = \frac{\sum_{i=1}^n p(j|i) (x_i + \rho_i \operatorname{sgn}(y_j - x_i)) + \theta (y_{j+1} + y_{j-1})}{2\theta + \sum_{i=1}^n P(j|i)}$$

where, the association probability distribution is now given by $p(j|i) = \frac{e^{-\beta d(x_i, y_j, \rho_i)}}{\sum_{k=1}^n e^{-\beta d(x_i, y_k, \rho_i)}}$ and $\operatorname{sgn}(\cdot)$ is a vector-valued signum function.

with the help of this equation find the distance of minimum cities at runtime. But it's difficult to find the solution manually for complex data sets even to difficult to implement it on runtime. This can be explained clearly in the figure

7 RUNTIME SOLUTION AND LENGTH OF TOUR BETWEEN CITIES

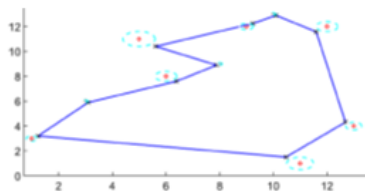


figure:4 from [2] Returning CESTP total length=14.57, Run time=6.82s

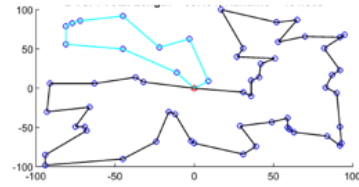


figure:5 from [2] 2-TSP total length=637.34, Run time=104.82

The MEP approach is purposed to solve the problem of SA and DA algorithms which having the solution for 59 number of cities optimal at run time but not compatible for the large datasets at run time so that it can make the comparison between them to get the sufficient and reliable results.

$$\min_{y_j \in \Omega, 1 \leq j \leq K} \sum_{i=1}^n \left\{ \min_{y_j \in \Omega, 1 \leq j \leq K} d(x_i, y_j) \right\}, \quad (1)$$

where $d(x_i, y_j) \in \mathbb{R}_+$ denotes the distance between the i^{th} city location x_i and j^{th} facility location y_j , and $\Omega \subset$

In the MEP follows this equation to get the reliable results. And the comparison of SA and DA demonstrated in the graph notation.

8 comparison of SA and DA representation

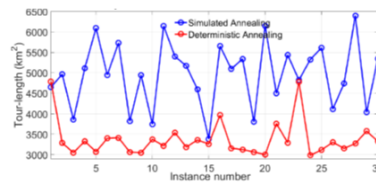
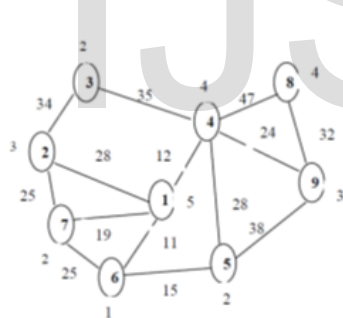


figure:6 from [2] A new approach based a genetic algorithm for the Selective Travelling Salesman Problem [3]. This research paper based on the genetic algorithm approach which has been concerned about to solve the shortest path problem and focused on that using minimized cost and total profit maximized. In which approach follows

the population of generation and optimized combination of cities between the length of shortest path tour. The algorithm code is such that

$$\text{Maximum } \sum_i p_i \sum_j x_{ij}$$

9 Graph representation of a Network of cities



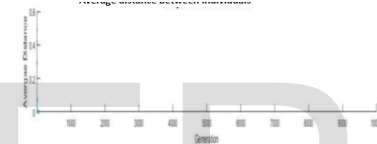
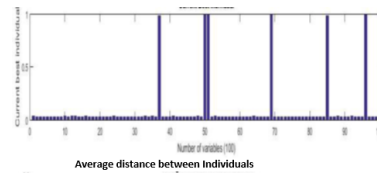
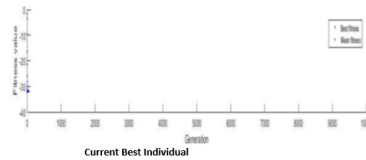
And the population gain and profit calculated by this algorithm

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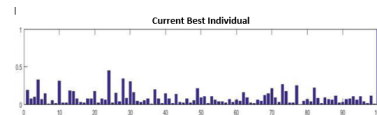
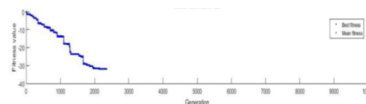
begin
I = {v1}, n:=m :=1 ;
begin
find v1 and vj such that
p1/c1r = max {pk /ckr} and p1/c1r = max {pk /ckr} ;
if (p1 /c1r > p1/c1r ) then add v1 to I before r and set r:= i
else add vj to I after s and set s:=j
end
close the current path by adding arc ( v1,vn )
end
end
    
```

With the help of this algorithm gets the best result of maximum cost ratio between gain and cost for using the same piece of code.

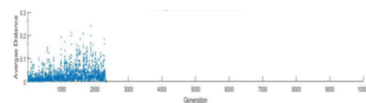
10 Results of initial Population



11 Results without Considering Initial Population



12 Average Distance between Individual



13 GELS-GA

Hybrid Metaheuristic Algorithm for Solving Multiple Travelling Salesman Problem[4].This paper has included new heuristic algorithm approach which is the type of NP complete problem solving.It is simple and efficient then two other algorithm approaches in which includes ACO and GA.Through this algorithm easily find the optimal solution for mTSP with the help of this equation

$$F(x) = \sum_{i=1}^n \sum_{j=2}^n \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2} + \sum_{i=1}^n \sum_{j=2}^n \text{cost}(c_i, c_j)$$

GELS has provides the best solution as compared to ACO and GA at the distance cover between multiple number of cities at minium cost.This minimized ratio of calculations represent in this figure Proposed method of algorithm representation

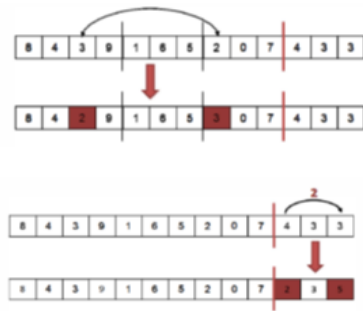


figure:10 [4] Explain the first part about proposed method for transfer mutation but the second part at left side describes about the salesman of mutation for proposed method

14 Comparison table of ACO and GELS-GA

Table: 1[4] Comprison show the result of Minimized Total Distance covered by different algorithms

Problem	ACO in [21]				GELS-GA				%Improvement of GELS-GA over	
	n	m	l	Avg. Time (s)	Best	Avg. Time (s)	Best	Worst	ACO	GELS-GA
g76	10	3	20	180000	31	17502	19	12184	3	24.51
g124	120	3	40	150000	128	13864	44	10505	3	24.84
g126	220	3	40	170000	141	14815	22	12113	9	10.97
g129	200	3	30	80000	208	8219	67	7654	11	8.9
g140	400	3	100	160000	563	12657	55	14022	16	11.22
g160	1000	3	200	300000	2628	34008	186	24244	27	6.6
								Minimum	8.48	0.86
								Mean	17.89	11.97
								Maximum	24.51	17.81

Comparison Show the results for Minimized travelled distance an Time by Salesman

Problem	Enhanced GA [24]					GELS-GA				%Improvement of GELS-GA over Enhanced GA	
	Name	n	Best	Avg.	Worst	Best	Avg.	Worst	Time (s)	Enhanced GA	GELS-GA
nTSP	g1	447.42	488.3	449.62	7.10	232	254.3	251	5	41.52	
	g2	436.11	478.51	482.61	8.78	220	262.5	260	6	41.65	
	g3	326.62	377.97	368.8	11.29	324	328	332	9	44.15	
mTSP	g1	22462.97	24464.41	22611.52	17.27	17300	18035	18270	12	19.92	
	g2	22862.48	24655.47	24564.96	14.18	18212	20080	20068	14	14.45	
	g3	22487.48	24053.53	24215.04	16.27	22534	24083.5	23245	17	11.57	
mMSP	g1	2066.42	2075.48	2075.05	1.48	1812	1868	2006	25	10.69	
	g2	16170.41	19511.04	18557.45	28.04	17040	17933.5	18287	25	1.62	
	g3	44213.12	46663.57	44663.57	51.62	38133	39355.5	38661	28	3.72	
nTSP	g1	44088.29	44528.57	44782.05	44.23	41113	41722	42263	32	6.39	
	g2	19326.26	24216.66	24753.87	19.73	17100	17440	17669	36	1.87	
	g3	31622.32	31808.99	31923.38	19.27	66474	66824.3	67113	44	0.84	
								Minimum	1.62		
								Mean	18.57		
								Maximum	44.15		

All of the calculations show that how much improvement has been discovered through GELS method of algorithm. These results are optimal and even mostly accepted in high complex data set. Estimation of Cost Rate Percentage for Travelling Salesman Problem Using Hybridized ACO Algorithm[5].ACO algorithm purposed to solve the shortest path and cover all data set at minimum cost. For such purpose includes equations rather than their algorithms like hamming and Euclidean distance standard equation in ACO.

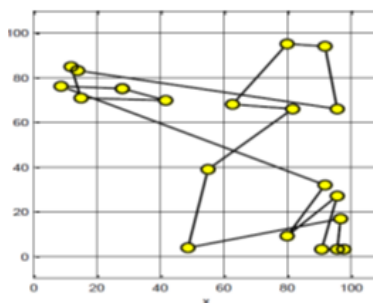
15 ALGORITHM FOR ACO

```

Procedure ACO_MetaHeuristic
while(not_termination)
generateSolutions()
daemonActions()
pheromoneUpdate()
end while
end procedure
    
```

Here are presented three equations with their results and formula.

17 TSP tour graph for visited node with equations



16 REPRESENTATION OF EQUATIONS

Eucliden equation

$$Dist_{Eucliden}((x, y), (a, b)) = \sqrt{(x - a)^2 + (y - b)^2} \quad (3)$$

Chebyshev equation

$$Dist_{Chebyshev}(p, q) = \max(|p-rq_i|)$$

Hamming equation

$$Dist_{Hamming}(i, j) = pdist2(p, q, 'hamming');$$

with the help of these three equations easily make comparison for ACO algorithm. They are not dependent on each other every equation has its own parallel data set of execution.

18 COST ESTIMATION ANALYSIS RESULTS OF EQUATIONS

So the graph representation of the equations describes about the different cases of equations with compare the other equations. Table:3 [5] Comparative study of Euclidean, Hamming and Chebyshev Distance equation[5]

S No.	Euclidean distance		Hamming distance		Chebyshev Distance	
	Best cost	Worst cost	Best cost	Worst cost	Best cost	Worst cost
Case 1:	410.55	622.51	20	20	278	368
Case 2:	462.85	703.12	20	20	280	486
Case 3:	428.37	428.37	20	20	244	356
Case 4:	14.70	683.18	20	20	300	404
Case 5:	685.18	423.37	20	20	276	354
Case 6:	371.53	524.03	20	20	212	240

In this table of comparison clearly shows that best results of three equation with the distance formula and also express the worst cost. Five-elements Cycle Optimization Algorithm for The Travelling Salesman Problem[6]. This algorithm approach has been developed to solve the multiple Travelling salesman problem and find the optimal solution for them. The five cycle model

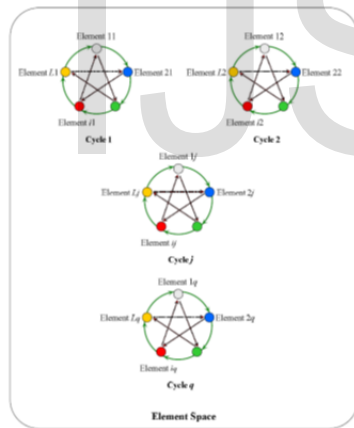
has been focused on generation and restriction of the worlds elements. But the FECO has been focused on the optimization of TSP.

19 Algorithm for FECO

$$F_{ij}(k) = \ln[m_{(i-1)j}(k)/m_{ij}(k)] - \ln[m_{(i-2)j}(k)/m_{ij}(k)] - \ln[m_{ij}(k)/m_{(i+1)j}(k)] - \ln[m_{ij}(k)/m_{(i+2)j}(k)] \quad [6]$$

After constructing Element Space, FECO can be established to solve TSP. To illustrate clearly, the corresponding relationship among FECM, FECO and TSP is shown in Table I.

20 STRUCTURE OF FECO ELEMENTS SPACE



21 Results for the Optimization comparison of FECO and TSP

Table 4[6]: Optimal results for FECO for 25 TSP instances

Instance	Optimal solution	Best	Worst	Average	D _{opt} (%)	StdDev
Instance1	3018785	3018785	3018785	3018785	0.0000	0.0000
Instance2	743097	743097	743097	743097	0.0000	0.0000
Instance30	4237466	4237466	4237466	4237466	0.0000	0.0000
Instance1	426	4283816	4324802	4305293	1.0632	1.1319
Instance2	7442	7444369	7444369	7444369	0.0314	0.0000
Instance30	426	4271096	4407941	4361111	0.4713	1.1313
Instance76	838	5404968	5576029	5444552	3.0586	1.8227
Instance99	1211	12314802	12519871	12424648	2.6447	5.5881
Instance101	629	4484409	4583476	4544567	1.0472	2.8841
Instance105	14379	14422959	14547844	14494258	0.8067	49.9560
Instance127	114252	1195091745	1214843358	1209514754	1.9542	466.0055
Instance130	4110	61392603	62424451	61974842	1.4318	22.5884
Instance139	6528	66134678	67131174	6661732	2.1328	39.1399
Instance150	20524	206191154	21372144	27903794	3.8888	116.6051
Instance159	20139	20301449	20814199	20448241	1.875	78.6062
Instance179	42889	421627497	430739730	427538693	1.5489	359.1648
Instance200	29568	298684254	30334597	300754627	2.3687	92.5010
Instance200	29437	300254271	304514723	302534533	2.5741	109.0839
Instance226	80569	809067530	808837667	807923345	0.4148	105.9287
Instance244	80135	80191466	81824622	81072688	3.9122	206.676
Instance299	48193	496911408	503594443	500184782	3.7922	181.0639
Instance318	43029	435234914	44042308	43944861	1.6552	178.6412
Instance400	17281	158899121	162442042	161278661	5.5367	63.1528
Instance442	50778	542757530	551093559	547484180	7.8854	177.8897
Instance575	6793	73062662	73612669	73483975	8.3803	12.5793

Table 4[6] Comparison results and Statistical Results for 6 Algorithms

Instance	Optimal solution	GA	ACO	Meta-GA	SSA	BSA	FECO
Instance1	3018785	3018785	3018785	3018785	3018785	3018785	3018785
Instance2	743097	743097	743097	743097	743097	743097	743097
Instance30	4237466	4237466	4237466	4237466	4237466	4237466	4237466
Instance1	426	4283816	4324802	4305293	4305293	4305293	4305293
Instance2	7442	7444369	7444369	7444369	7444369	7444369	7444369
Instance30	426	4271096	4407941	4361111	4361111	4361111	4361111
Instance76	838	5404968	5576029	5444552	5444552	5444552	5444552
Instance99	1211	12314802	12519871	12424648	12424648	12424648	12424648
Instance101	629	4484409	4583476	4544567	4544567	4544567	4544567
Instance105	14379	14422959	14547844	14494258	14494258	14494258	14494258
Instance127	114252	1195091745	1214843358	1209514754	1209514754	1209514754	1209514754
Instance130	4110	61392603	62424451	61974842	61974842	61974842	61974842
Instance139	6528	66134678	67131174	6661732	6661732	6661732	6661732
Instance150	20524	206191154	21372144	27903794	27903794	27903794	27903794
Instance159	20139	20301449	20814199	20448241	20448241	20448241	20448241
Instance179	42889	421627497	430739730	427538693	427538693	427538693	427538693
Instance200	29568	298684254	30334597	300754627	300754627	300754627	300754627
Instance200	29437	300254271	304514723	302534533	302534533	302534533	302534533
Instance226	80569	809067530	808837667	807923345	807923345	807923345	807923345
Instance244	80135	80191466	81824622	81072688	81072688	81072688	81072688
Instance299	48193	496911408	503594443	500184782	500184782	500184782	500184782
Instance318	43029	435234914	44042308	43944861	43944861	43944861	43944861
Instance400	17281	158899121	162442042	161278661	161278661	161278661	161278661
Instance442	50778	542757530	551093559	547484180	547484180	547484180	547484180
Instance575	6793	73062662	73612669	73483975	73483975	73483975	73483975
z-score		1.2801	1.1801	1.2801	1.2801	1.2801	1.2801

ALGORITHM TO SOLVE TRAVEL SALESMAN PROBLEM EFFICIENTLY[7]. This research paper has been design to solve the problem with efficiently an optimally solutions. In this paper discuss the many previous techniques uses as TSP solution finding but focus on the new purposed method of optimal and efficient results are Pick near Point(PNP).

22 Algorithm for PNP

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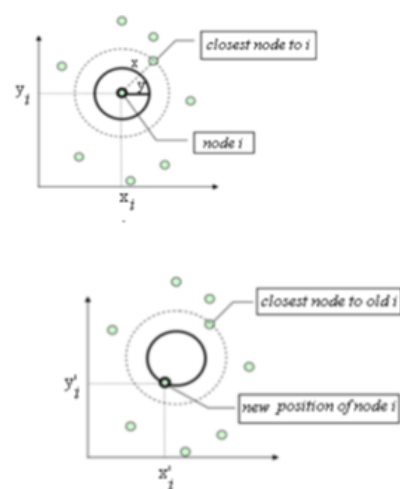
Algorithm 1 Algorithm PNP
Input: C // XY coordinates of cities.
Output: Path // Set of edges
Require: dbest // Distance Matrix calculate using Eq. 1
Require: S4 // Specific Distance calculate using Eq. 2
Path // Maximum cost rectangle by use of corner edges
while iterate == True do
    Cities = C - Path
    nearPoints = nearPoints(Cities, Path, S4)
    for all nearPoints do
        for all Path do
            Dist // minimum distance from path to point
            if minDist > Dist then
                minDist = Dist
            else
                minDist = minDist
            end if
        end for
    end for
end while

function nearPoints(Cities, Path, S4)
    for all Path do
        for all Cities do
            Distance // Min distance from nearest point
            if Distance < S4 then
                nearPoint = City
            end if
        end for
    end for
end function
    
```

This algorithm of PNP express the large

data set instances for TSP with efficiently results getting. With the help of such algorithm easily can be find the solution for 100 number of cities with minimum distance travelled very efficiently. But this is also an approximate algorithm as compare to the another algorithms but not consider as exact. Adaptability of a Discrete PSO Algorithm applied to the Traveling Salesman Problem with Fuzzy Data[8]. In this paper describes about the behaviour of optimal solutions by using Particle swarm Optimization implementations on TSP uncertain environment. It is going to follow that uncertain or fuzzy data by discrete PSO equation[8].

23 Initial map and its fuzzied map graphical representation



fuzzie form figure:13[8] In this diagram[8] has been explain about the

fuzzines of data set through PSO algorithm finding their behaviour. Which can be clearly explain that how much different after them.

24 Comparison of Fuzzied data with other algorithms

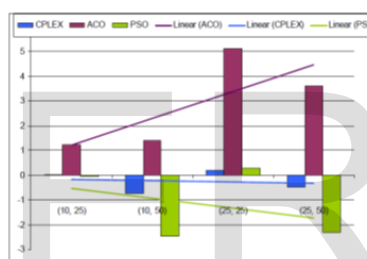


figure:14[8] In which graph describes about the other algorithm at the level of percentage error as compare to the PSO algorithm. That having the Path Planning and Following using Genetic Algorithms to Solve the Multi-Travel Salesman Problem in Dynamic Scenario[9]. In this paper having the algorithm about path planning of many number of cities distributed randomly. Here we are going to generate the best path for robots through this algorithm for dynamic set of problems. With the help of Genetic algorithm easily find the MTSP best solutions efficiently. In this paper used algorithm has concept about the chromosomes and population. With the help of this concept easily can be found the solution of problems with best and optimal results.

25 Process of Genetic Algorithm

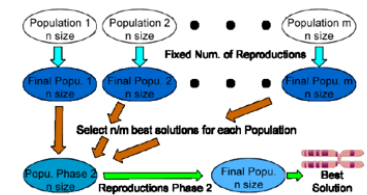


figure:15[9] genetic algorithm process flow representation In which the figure[9] explain about the population how to reproduce through the chromosome and how much increase the number of population. Now after to clearly define the genetic algorithm how to solve the MTSP here are discuss about the path planning for robots. In which focused on this agent has been move through the shortest and optimal path for particular scenarios.

26 Path Planning Representation

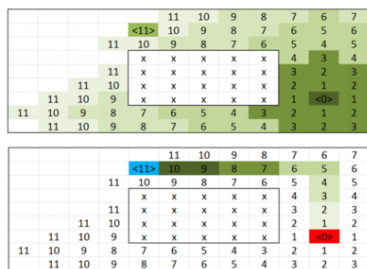


figure:16[9] In this figure express about to find the shortest path for robots agent has been visited each place with the minimum distance. This can be held through with the algorithm to achieve it. So with the help of this algorithm we can achieve best results for path planning through Ga algorithm

approach.

27 Results of GA

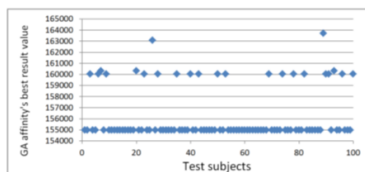


figure :17(a)

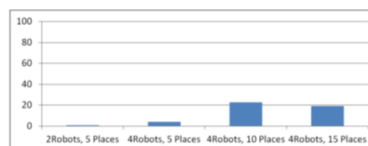


figure:17(b) IN which figure:17[9] a explain about the dispersion of best results though the algorithm of GA but the second part for Figure:17(b) define about the percentage error of GA with respect to best results.

28 Comparison

The Stochastic Traveling Saleman TSP in which has to cover the trajectory length of shortest path for n number of distributed cities. Those trajectory having the high probabilities for orienteering Problem for these problems which are different then the other approaches. This advantage is one of unique then others. Maximum Entropy Principle based Approach for MTSP in the paper focus on the run time marginal increase for simulated annealing algorithm. For the solution of mTSP using CETSP that is computational challenge for it. So with this proves the best comparison results provide in the SA and DA algorithm which is best working of it. A new approach of GA for Selective TSP al-

gorithm is the heuristic approach following though the help of this easily can be achieved the optimal combination of cities. That is the best feature of it rather than others. GELSGA heuristic approach for MTSP this approach has been different then the Maximum Entropy Principle that also used to solve multiple problems. This is the best as compared to ACO and GA algorithm also. Path Planning an following for MTSP it can be focused on the best path finding for robots agents with the minimum cost at total travel distance covered each place. This is the advantage of ACO algorithm for path planning which has been unique to others. Harmony search algorithm for TSP in which having the algorithm for the solving the optimal problem results the traveling salesman according to artificial intelligence field. Which is the unique of this algorithm form other that can be covered only in this paper.

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