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| LITERATURE REVIEW 1.0  Global Optimization Using Meta-Heuristics | |  |  | | --- | --- | | Faiza Shanawar | 15140070 | | Haider Ali | 15140101 | | Mohsin Qamar | 15140104 | | Usama Imran | 15140098 | |

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***Document Information***

|  |  |
| --- | --- |
| User |  |
| Project | Meta-Heuristics for Global Optimization |
| Document Version | 1.0 |
| Document ID | AR-02 |
| Identifier | Literature Review 1.0 |
| Status | Draft |
| Authors(s) | Usama Imran |
| Approver(s) | Syed Qamar Askari |
| Issue Date |  |

# Introduction:

## Mathematical Optimization:

Mathematical Optimization can be defined as a technique of finding a maximum or minimum value of a function of several variables subject to set of constraints. More generally, optimization includes finding "best available" values of some objective function given a defined domain (or input), including a variety of different types of objective functions and different types of domains.



Graph of a paraboloid given by

z = f(x, y) = −(x² + y²) + 4. The global maximum at (x, y, z) = (0, 0, 4) is indicated by a blue dot.

### Types:

Optimization is classified into following types:

* Discrete optimization
* Continues optimization
* Single Objective Optimization
* Multi Objective Optimization
* Many Objective Optimization
* Combinatorial optimization
* Unconstrained optimization
* Constrained optimization

### Applications:

Optimization techniques are being used in real world problems some of their applications are as follows:

* Feature Selection
* Automatic Clustering
* Time Scheduling
* Wireless Sensor Network Optimization
* Vehicle Routing Problem
* Watermarking
* Bioinformatics
* Circuit Designing
* Game Strategy Planning
* Power Supply Management

## Meta-Heuristics:

Heuristic is a Greek word which means “to solve”. It pertains to trial-and-error method of problem solving used when an exact algorithmic approach is impractical. Main characteristic of meta-heuristics is that they are problem independent. Meta-heuristics give us a way to solve complex problems that are not solvable in polynomial time (NP-Hard Problems). Although they don’t give exact solution of a particular problem, meta-heuristics provide guidelines that can give best solution available

## Properties of Meta-Heuristic Algorithms:

## Related Work:

Following are the lists of optimization Algorithms:

|  |  |
| --- | --- |
| **Algorithm** | **Inspiration** |
| Genetic Algorithm | This algorithm is inspired by biological processes: crossover and mutation. |
| Gray Wolf Optimization | This algorithm is inspired by pack of wolves that are search for hunt. |
| Particle Swarm Optimization | This algorithm is inspired by flock of birds. |
| Gravitational Search Algorithm | This algorithm is inspired by Newton’s Law of Gravitation. |
| Whale Optimization | This algorithm is inspired by Whale hunting. |
| Dragonfly Optimization | This algorithm is inspired by dragonflies and their hunting behavior. |
| Centipede Optimization | This algorithm is inspired by centipede’s hunting behavior |
| Bat Optimization | This algorithm is inspired by bat’s echolocation |
| Ant Colony Optimization | This algorithm is inspired by ant’s colony |
| Artificial Bee Colony Optimization | This algorithm is inspired by Bees |
|  |  |

## Benchmark Functions:

After the algorithms are proposed, they are tested on some evaluation functions which are known as benchmark functions. Algorithms are subjected to benchmark functions their performance is checked through comparison with already existed algorithms.

### Types:

Following are the types of benchmark functions:

1. Unimodal
2. Multimodal
3. Differentiable
4. Non-differentiable
5. 1- Dimensional
6. 2-Dimentional
7. 3-Dimensoinal
8. N-Dimensional
9. Convex
10. Non-convex
11. Parametric
12. Separable

### Overview:

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| ***Function name*** | ***Graph*** | ***Equation*** | ***Nature*** | ***Range*** |
| Ackley N. 2 Function |  | f(x,y)=−200e−0.2√x2+y2 | N-dimensional, unimodal, convex, differentiable | [-6,6] |
| Booth Function |  | f(x,y)=(x+2y−7)2+(2x+y−5)2 | 2-dimensional, continuous, convex, differentiable, non-separable, unimodal convex | [−10,10] |
| Brent Function |  | f(x,y)=(x+10)2+(y+10)2+e−x2−y2 | 2-dimensional continuous convex differentiable non-separable unimodal convex | [−10,10] |
| Drop-Wave Function |  | f(x,y)=−1+cos(12√x2+y2) (0.5(x2+y2)+2) | 2-dimensional continuous unimodal non-convex | [−5.2,5.2] |
| Exponential Function |  | f(x)=f(x1,...,xn)= −exp(−0.5n∑i=1x2i) | n-dimensional continuous differentiable non-separable unimodal convex | [−1,1] |
| Leon Function |  | f(x,y)=100(y−x3)2+(1−x)2 | 2-dimensional continuous differentiable non-separable unimodal non-convex | [0,10] |
| Deckkers-Aarts Function |  | (x,y)=105x2+y2−  (x2+y2)2+10−5(x2+y2)4 | 2-dimensional continuous differentiable non-separable multimodal non-convex | [−20,20] |
| Styblinski-Tank Function |  | f(x)=f(x1,...,xn)=  12n∑i=1(x4i−16x2i+5xi) | n-dimensional continuous multimodal non-convex | [−5,5] |
| Bartels Conn Function |  | f(x,y)=  |x2+y2+xy|+|sin(x)|+|cos(y)| | 2-dimensional non-separable multimodal non-convex non-differentiable | [−500,500] |
| Schwefel 2.20 Function |  | f(x)=f(x1,...,xn)=n∑i=1|xi| | 2-dimensional continuous differentiable non-separable unimodal non-convex | [−100,100] |
| Egg Crate Function |  | f(x,y)=x2+y2+25(sin2(x)+sin2(y)) | 2-dimensional continuous differentiable separable multimodal non-convex | [−5,5] |
| Shubert Function |  | f(x)=f(x1,...,xn)=  n∏i=1(5∑j=1cos((j+1)xi+j)) | n-dimensional, continuous, differentiable non-separable, multimodal, non-convex | [-10,10] |
|  |  |  |  |  |

# No Free Lunch Theorem:

# Literature Review:

## Genetic Algorithm:

### Inspiration:

Genetic Algorithm is one of the most famous optimization algorithms. It is inspired by the Darwin’s Theory of Evolution (natural selection). This algorithm exhibits the process of natural selection until the fittest population is obtained.



### Working:

This algorithm has five phases in which two are major phases: “Crossover” and “Mutation”. Algorithm starts from instantiating a random population and their fitness is calculated by **Objective Function**[[1]](#footnote-1)**.** Each member of population is named “**gene**”.This is 1st phase which can be called as “population initialization”. In the 2nd phase, fittest genes are selected which are to be subjected to crossover and mutation. This 2nd phase is called “**selection**”**.** After the selection of fittest genes, they undergo “**crossover”** phase which is 3rd phase of this algorithm. After crossover, genes go through “**mutation**” phase which is 4th phase of this algorithm. After mutation, fitness is calculated again and fittest genes after crossover and mutation are replaced by least fit genes in initialized population. This process is repeated until fittest population is obtained and then it is terminated in the last phase of this algorithm which is called “**termination**”.

1. Objective Function is basically a function that describes the fitness of each sample of population which needs to be either maximized or minimized. Every problem on which metaheuristic algorithms are applied, requires objective function which is minimized of maximized according to the need. [↑](#footnote-ref-1)