

```

import numpy as np
import math
import random
import matplotlib.pyplot as plt

def initialization(N,Dim,UB,LB):
    X=np.zeros((N,Dim))
    for i in range(N):
        for j in range(Dim):
            X[i,j]=random.random()*(UB-LB)+LB
    return X

def fitness(x):
    fitval=x[0] # will change beased on requirement
    return fitval

def AOA(N,M_Iter,LB,UB,Dim):
    # Two variables to keep the positions and the fitness value of the best-obtained solution
    Best_P=np.zeros((1,Dim))
    Best_FF=math.inf
    Conv_curve=[]
    # Initialize the positions of solution
    X=initialization(N,Dim,UB,LB)
    Xnew=X
    Ffun=np.zeros((1,X.shape[0])) # (fitness values)
    Ffun_new=np.zeros((1,Xnew.shape[0])) # (fitness values)

    MOP_Max=1;
    MOP_Min=0.2;
    C_Iter=1;
    Alpha=5;
    Mu=0.499;

    for i in range(X.shape[0]):
        Ffun=fitness(X[i,:]) # Calculate the fitness values of solutions
        if Ffun<Best_FF:
            Best_FF=Ffun
            Best_P=X[i,:]
    while C_Iter<M_Iter+1: # Main loop
        MOP=1-((C_Iter)**(1/Alpha)/(M_Iter)**(1/Alpha)); # Probability Ratio
        MOA=MOP_Min+C_Iter*((MOP_Max-MOP_Min)/M_Iter); # Accelerated function

        # Update the Position of solutions
        for i in range(X.shape[0]): # if each of the UB and LB has a just value
            for j in range(X.shape[1]):
                r1=random.random();
                if r1<MOA:
                    # Exploration phase
                    r2=random.random();
                    if r2>0.5:
                        Xnew[i,j]=Best_P[j]/(MOP+1)*((UB-LB)*Mu+LB)
                    else:
                        Xnew[i,j]=Best_P[j]*MOP*((UB-LB)*Mu+LB);
                else:
                    # Exploitation phase
                    r3=random.random();
                    if r3>0.5:
                        Xnew[i,j]=Best_P[j]-MOP*((UB-LB)*Mu+LB);
                    else:
                        Xnew[i,j]=Best_P[j]+MOP*((UB-LB)*Mu+LB);

        # Flag_UB=Xnew[i]>UB[0] # check if they exceed (up) the boundaries
        # Flag_LB=Xnew[i,:]<LB[0] # check if they exceed (down) the boundaries
        # Xnew[i,:]=(Xnew[i,:]*(~(Flag_UB+Flag_LB)))+UB*Flag_UB+LB*Flag_LB

        Ffun_new=fitness(Xnew[i,:]); # calculate Fitness function
        if Ffun_new<Ffun:
            X[i,:]=Xnew[i,:]
            Ffun=Ffun_new
        if Ffun<Best_FF:
            Best_FF=Ffun
            Best_P=X[i,:]
        Conv_curve.append(Best_FF)

    print('Iteration - ',str(C_Iter+1),': Best Position',str(Best_P),': Best Fitness',str("%.2f"%Best_FF))

    C_Iter=C_Iter+1; # incremental iteration

    return [Best_FF,Best_P,Conv_curve]

Solution no=25 # Number of search solutions

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M_Iter=100 # Maximum number of iterations
```

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LB=0
```

```
UB=1
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```
Dim=2
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[Best_FF,Best_P,Conv_curve]=AOA(Solution_no,M_Iter,LB,UB,Dim) # call the AOA
```

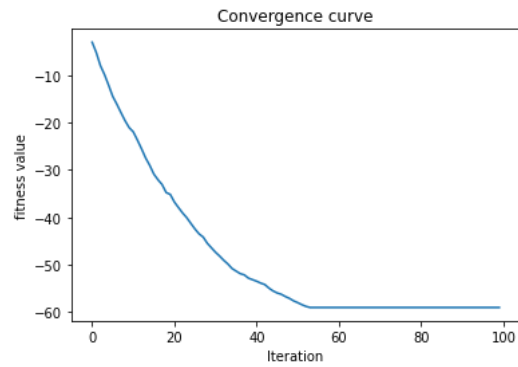
```
Iteration - 2 : Best Position [-2.98733212  0.15004695] : Best Fitness -2.99
Iteration - 3 : Best Position [-5.15377035  0.32736025] : Best Fitness -5.15
Iteration - 4 : Best Position [-7.92058632 -0.0246788 ] : Best Fitness -7.92
Iteration - 5 : Best Position [-9.81556652 -0.08850248] : Best Fitness -9.82
Iteration - 6 : Best Position [-12.06465797 -0.48495566] : Best Fitness -12.06
Iteration - 7 : Best Position [-14.42668965  0.15570514] : Best Fitness -14.43
Iteration - 8 : Best Position [-16.07332627  0.39395128] : Best Fitness -16.07
Iteration - 9 : Best Position [-17.85437768 -0.26309005] : Best Fitness -17.85
Iteration - 10 : Best Position [-19.57083406  0.20569426] : Best Fitness -19.57
Iteration - 11 : Best Position [-21.04405234 -0.39967414] : Best Fitness -21.04
Iteration - 12 : Best Position [-21.93451773 -0.17605538] : Best Fitness -21.93
Iteration - 13 : Best Position [-23.65911468  0.17076056] : Best Fitness -23.66
Iteration - 14 : Best Position [-25.49820682 -0.48074566] : Best Fitness -25.50
Iteration - 15 : Best Position [-27.44503423 -0.09197575] : Best Fitness -27.45
Iteration - 16 : Best Position [-29.02059964  0.43778755] : Best Fitness -29.02
Iteration - 17 : Best Position [-30.85804831 -0.45793725] : Best Fitness -30.86
Iteration - 18 : Best Position [-32.0492598  0.1535518] : Best Fitness -32.05
Iteration - 19 : Best Position [-33.06339359  0.47786391] : Best Fitness -33.06
Iteration - 20 : Best Position [-3.47557078e+01 -2.18409540e-02] : Best Fitness -34.76
Iteration - 21 : Best Position [-35.1677126 -0.12147359] : Best Fitness -35.17
Iteration - 22 : Best Position [-36.773175  0.07130415] : Best Fitness -36.77
Iteration - 23 : Best Position [-37.94654774  0.06812021] : Best Fitness -37.95
Iteration - 24 : Best Position [-39.0902941 -0.12977758] : Best Fitness -39.09
Iteration - 25 : Best Position [-40.08152361  0.12133421] : Best Fitness -40.08
Iteration - 26 : Best Position [-41.28981077 -0.09091793] : Best Fitness -41.29
Iteration - 27 : Best Position [-42.468317 -0.09575381] : Best Fitness -42.47
Iteration - 28 : Best Position [-43.50298217  0.0544152 ] : Best Fitness -43.50
Iteration - 29 : Best Position [-44.17593801  0.27873381] : Best Fitness -44.18
Iteration - 30 : Best Position [-45.48915571  0.12161844] : Best Fitness -45.49
Iteration - 31 : Best Position [-46.45021585 -0.10669354] : Best Fitness -46.45
Iteration - 32 : Best Position [-47.3880507  0.10431519] : Best Fitness -47.39
Iteration - 33 : Best Position [-48.20156305  0.10169547] : Best Fitness -48.20
Iteration - 34 : Best Position [-4.90946899e+01 -8.56640457e-04] : Best Fitness -49.09
Iteration - 35 : Best Position [-49.86942878 -0.08100722] : Best Fitness -49.87
Iteration - 36 : Best Position [-5.08144695e+01 -9.95087268e-05] : Best Fitness -50.81
Iteration - 37 : Best Position [-51.36778132  0.09188831] : Best Fitness -51.37
Iteration - 38 : Best Position [-5.19076820e+01  5.05873644e-06] : Best Fitness -51.91
Iteration - 39 : Best Position [-52.17107016 -0.08779587] : Best Fitness -52.17
Iteration - 40 : Best Position [-52.85630423  0.0826336 ] : Best Fitness -52.86
Iteration - 41 : Best Position [-53.19052803 -0.05319393] : Best Fitness -53.19
Iteration - 42 : Best Position [-53.51652481 -0.07255022] : Best Fitness -53.52
Iteration - 43 : Best Position [-53.91393575  0.08919467] : Best Fitness -53.91
Iteration - 44 : Best Position [-54.22394873 -0.06190007] : Best Fitness -54.22
Iteration - 45 : Best Position [-5.49795565e+01  3.26591842e-02] : Best Fitness -54.98
Iteration - 46 : Best Position [-55.56878308  0.07410811] : Best Fitness -55.57
Iteration - 47 : Best Position [-55.99945992 -0.07158681] : Best Fitness -56.00
Iteration - 48 : Best Position [-56.27921166 -0.07702045] : Best Fitness -56.28
Iteration - 49 : Best Position [-56.75610405 -0.06776491] : Best Fitness -56.76
Iteration - 50 : Best Position [-5.71541858e+01 -1.69093143e-06] : Best Fitness -57.15
Iteration - 51 : Best Position [-5.76709479e+01  2.87104075e-02] : Best Fitness -57.67
Iteration - 52 : Best Position [-5.80481763e+01 -1.23376896e-02] : Best Fitness -58.05
Iteration - 53 : Best Position [-5.84763966e+01  3.69612813e-03] : Best Fitness -58.48
Iteration - 54 : Best Position [-5.88334158e+01 -3.11907536e-07] : Best Fitness -58.83
Iteration - 55 : Best Position [-5.90648442e+01  2.99363897e-03] : Best Fitness -59.06
Iteration - 56 : Best Position [-5.90086090e+01  1.68347897e-04] : Best Fitness -59.06
Iteration - 57 : Best Position [-2.65394288e+01  7.57153421e-05] : Best Fitness -59.06
Iteration - 58 : Best Position [-2.65924897e+01  3.41505704e-05] : Best Fitness -59.06
Iteration - 59 : Best Position [-1.20281037e+01  1.54467147e-05] : Best Fitness -59.06
```

```
Conv_curve=np.array(Conv_curve)
print("\nBest solution found:\n")
print('Best fitness :',Best_FF)
print('Best position :',Best_P)
plt.plot(Conv_curve)
plt.xlabel('Iteration')
plt.ylabel('fitness value')
plt.title('Convergence curve')
plt.show()
```

Best solution found:

Best fitness : -59.064844193054405

Best position : [5.00006080e-04 6.68432893e-14]



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