▼ Defferential Evolution (Sphere Func.)

```
import numpy as np
import random
from array import *
import matplotlib.pyplot as plt
import statistics as st
import math
D = 10 # Dimensi dari permasalahan
maxit = 99 #Max iterasi
ukuranPopulasi = 10
LB = -10 #Batas Bawah
UB = 10 #Batas Atas
CR = 0.9
F = 0.5
import random
def initPopulasi(ukuranPopulasi, D, UB, LB): #Inisialisasi setiap baris kolom
    P = np.empty((ukuranPopulasi,D))
   for i in range(ukuranPopulasi):
        for j in range(D):
            P[i][j]= (random.random()*(UB-LB))+LB
    return P
P = initPopulasi(ukuranPopulasi, D, UB, LB)
print(P)
     [[ 5.69622364 -9.77272772 -8.23100607 -6.47651935 -7.01127811 2.86234108
        9.15963831 -8.82896704 -6.82454823 -8.44254531]
      [-8.50629736 6.44187377 2.43136056 -5.28877532 1.70407556 6.15200632
       -0.80228779 -1.23550781 -1.64848515 -0.67735818]
      [ 3.96002675 -2.56211483 -1.17381326 4.28349931 -1.97867226 2.51688329
        7.15746641 -4.11080985 2.21996604 -7.81933746
      [ 2.66710187    4.31371608    -3.87726489    5.30822543    8.99155279    -2.91639314
        7.78476023 4.32537916 0.38064641 4.77324052]
      [ 4.96449287 -3.61168306  9.56688097  3.48700723  5.9448273  2.48740972
        9.05995896 -1.06552936 6.83874003 8.77041278]
      [-4.85051854 -7.43523641 -6.37721882 -8.45204101 9.24913917 -0.03213903
       -5.86917081 -9.92411394 -6.26045077 9.69013618]
      [-0.34018508 8.72744673 -4.44127703 -8.22533886 -2.42806967 8.74088018
       -3.05250204 -7.61924065 5.51878122 1.66888765]
      [ 0.78397727 -5.63639414 -4.88399924 -3.32307591 -4.88506801 2.59119844
        5.51257767 -1.91504583 -1.91480962 6.93258055]
      [-9.02182498 8.99499721 -5.34738778 4.37025215 -4.30535137 5.39403455
        2.56665363 7.14115949 -4.09935427 -4.58248127]
                                6.77122162 -6.75696552 -6.86499639 -2.38402261
      [-1.9732305 7.363828
       -5.90980649 0.24289997 -6.75298643 5.1915824 ]]
```

```
ukuran = populasi.shape
    ukuranPopulasi = ukuran[0]
    dimensi = ukuran[1]
    Fobj = np.empty(ukuranPopulasi)
    for i in range (ukuranPopulasi):
        d = 0
        for j in range (dimensi):
            d = d + populasi[i][j]**2
            # d = d + np.sum(np.abs(populasi[i][j])) + np.product(np.abs(populasi[i][j]))
            \# d = d + np.sum(100*(populasi[i][j]+1)-(populasi[i][j]))**2 + (populasi[i][j])
            \# d = d + np.sum((populasi[i][j]**2) - (10*math.cos(2*math.pi*populasi[i][j])
        Fobj[i] = d
    return Fobj
Fobj = funcSphere(P)
print(Fobj)
     [574.69961469 193.83500895 186.42066406 260.52550295 389.8097685
      542.49098258 346.57643084 183.64112218 353.01632961 309.97753267
def crossOver(populasi,F,j):
    ukuran = populasi.shape
    ukuranPopulasi = ukuran[0]
    Dimensi = ukuran[1]
    r1 = random.randint(0, ukuranPopulasi-1)
    r2 = random.randint(0, ukuranPopulasi-1)
    r3 = random.randint(0, ukuranPopulasi-1)
    while r1 == r2:
        r2 = (r2 + 1)%ukuranPopulasi
    while r3 == r1 or r3 == r2:
        r3 = (r3 + 1)%ukuranPopulasi
    V = P[r3][j] + F*(P[r1][j] - P[r2][j])
    return v
v = crossOver(P,F,1)
print(v)
     0.23808210577046385
P = initPopulasi(ukuranPopulasi, D, UB, LB)
print("Inisialisasi Populasi :\n",P)
print(P)
Fobj = funcSphere(P)
print(Fobj)
```

```
print("\nFungsi Objective :\n",Fobj)
U = np.empty((1,D))
bestFobj = np.empty((maxit+1))
bestFobj[0] = Fobj.min()
for it in range(maxit):
    for i in range(ukuranPopulasi):
        for j in range(D):
            U[0][j] = P[i][j]
        jrand=random.randint(0,D)
        for j in range(D):
            if random.random() < CR or j == jrand:</pre>
                v = crossOver(P,F,j)
                U[0][j] = v
        FobjU = funcSphere(U)
        if FobjU < Fobj[i]:</pre>
            Fobj[i] = FobjU
            for j in range(D):
                P[i][j] = U[0][j]
    bestFobj[it+1] = Fobj.min()
    print("\nNilai Optimal : \n",bestFobj)
    for i in range(30):
        bestFobj
      7.64587745e+00 7.64587745e+00 5.67610107e+00 5.28473509e+00
      4.53707649e+00 3.58652646e+00 3.58652646e+00 1.06292963e+00
      1.06292963e+00 1.06292963e+00 1.06292963e+00 1.06292963e+00
      9.12842930e-01 9.12842930e-01 9.12842930e-01 7.81364918e-01
      5.88571546e-01 3.86149953e-01 3.86149953e-01 3.86149953e-01
      3.86149953e-01 3.86149953e-01 3.86149953e-01 2.79320565e-01
      2.35460975e-01 2.35460975e-01 2.31905451e-01 1.95616440e-01
      1.95616440e-01 1.95616440e-01 1.49883717e-01 1.49883717e-01
      1.49883717e-01 1.26493585e-01 1.02858042e-01 8.93648664e-02
      3.19110875e-02 3.19110875e-02 3.19110875e-02 1.75648542e-02
      1.75648542e-02 1.75648542e-02 1.75648542e-02 1.75648542e-02
      1.75648542e-02 1.75648542e-02 1.75648542e-02 1.75648542e-02
      1.07193586e-02 1.07193586e-02 1.07193586e-02 1.07193586e-02
      9.60403792e-03 5.60006056e-03 5.15511046e-03 3.32847529e-03
      3.32847529e-03 2.94986216e-03 4.01274509e+00 9.68362150e+00]
     Nilai Optimal :
      [2.58455405e+02 2.58455405e+02 2.58455405e+02 2.58455405e+02
      1.79098417e+02 1.66839068e+02 1.66839068e+02 1.61870264e+02
      1.61870264e+02 1.57014684e+02 1.17593958e+02 1.17593958e+02
      1.01152373e+02 1.01152373e+02 1.01152373e+02 9.48490093e+01
      9.48490093e+01 8.30231256e+01 8.30231256e+01 8.30231256e+01
      8.10216226e+01 8.10216226e+01 8.10216226e+01 8.10216226e+01
      7.39386141e+01 6.31351088e+01 6.31351088e+01 6.31351088e+01
      4.23278480e+01 4.10339568e+01 4.10339568e+01 1.75357761e+01
      1 75357774...01 1 75303533...01 1 75303533...01 1 75303533...01
```

```
1./535//610+01 1./52825230+01 1./52825230+01 1./52825230+01
      1.75282523e+01 1.01260533e+01 1.01260533e+01 1.01260533e+01
      7.64587745e+00 7.64587745e+00 5.67610107e+00 5.28473509e+00
      4.53707649e+00 3.58652646e+00 3.58652646e+00 1.06292963e+00
      1.06292963e+00 1.06292963e+00 1.06292963e+00 1.06292963e+00
      9.12842930e-01 9.12842930e-01 9.12842930e-01 7.81364918e-01
      5.88571546e-01 3.86149953e-01 3.86149953e-01 3.86149953e-01
      3.86149953e-01 3.86149953e-01 3.86149953e-01 2.79320565e-01
      2.35460975e-01 2.35460975e-01 2.31905451e-01 1.95616440e-01
      1.95616440e-01 1.95616440e-01 1.49883717e-01 1.49883717e-01
      1.49883717e-01 1.26493585e-01 1.02858042e-01 8.93648664e-02
      3.19110875e-02 3.19110875e-02 3.19110875e-02 1.75648542e-02
      1.75648542e-02 1.75648542e-02 1.75648542e-02 1.75648542e-02
      1.75648542e-02 1.75648542e-02 1.75648542e-02 1.75648542e-02
      1.07193586e-02 1.07193586e-02 1.07193586e-02 1.07193586e-02
      9.60403792e-03 5.60006056e-03 5.15511046e-03 3.32847529e-03
      3.32847529e-03 2.94986216e-03 2.94986216e-03 9.68362150e+00]
     Nilai Optimal:
      [2.58455405e+02 2.58455405e+02 2.58455405e+02 2.58455405e+02
      1.79098417e+02 1.66839068e+02 1.66839068e+02 1.61870264e+02
      1.61870264e+02 1.57014684e+02 1.17593958e+02 1.17593958e+02
      1.01152373e+02 1.01152373e+02 1.01152373e+02 9.48490093e+01
      9.48490093e+01 8.30231256e+01 8.30231256e+01 8.30231256e+01
      8.10216226e+01 8.10216226e+01 8.10216226e+01 8.10216226e+01
      7.39386141e+01 6.31351088e+01 6.31351088e+01 6.31351088e+01
      4.23278480e+01 4.10339568e+01 4.10339568e+01 1.75357761e+01
      1.75357761e+01 1.75282523e+01 1.75282523e+01 1.75282523e+01
      1.75282523e+01 1.01260533e+01 1.01260533e+01 1.01260533e+01
      7.64587745e+00 7.64587745e+00 5.67610107e+00 5.28473509e+00
      4.53707649e+00 3.58652646e+00 3.58652646e+00 1.06292963e+00
      1.06292963e+00 1.06292963e+00 1.06292963e+00 1.06292963e+00
      9.12842930e-01 9.12842930e-01 9.12842930e-01 7.81364918e-01
import statistics as st
print("Nilai Mean : ",st.mean(bestFobj))
print("Nilai Standard Deviation : ",st.stdev(bestFobj))
print("Nilai Minimal : ",np.min(bestFobj))
    Nilai Mean : 39.045273875157235
     Nilai Standard Deviation: 65.73583350532162
     Nilai Minimal: 0.00294986216248811
x = np.linspace(0, 1, maxit+1)
plt.plot(x, bestFobj, label= 'f=Fobj')
plt.show()
```



Genetic Algorithm

```
import numpy as np
import random
from array import *
import matplotlib.pyplot as plt
import statistics as st
import math
N = 3
rows, cols = (N, 4)
induk = [[0 for i in range(cols)] for j in range(rows)]
print (induk)
N_Anak = 6
rows, cols = (N_Anak, 4)
anak = [[0 for i in range(cols)] for j in range(rows)]
print(anak)
     [[0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0]]
     [[0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0], [0, 0, 0, 0]]
#Inisialiasi Parameter
Pc = 0.9
Pm = 0.1
for i in range(N):
    a = random.randint(0, 15)
    induk[i] = [int (x) for x in '{:04b}'.format(a)]
    print(a, induk[i])
     15 [1, 1, 1, 1]
     0 [0, 0, 0, 0]
     6 [0, 1, 1, 0]
#Function cross over
def crossover(ind1=[], ind2=[]):
    anak1 = [0, 0, 0, 0]
```

anak2 = [0, 0, 0, 0]

```
anak1[0] = ind1[0]
    anak1[1] = ind1[1]
    anak1[2] = ind2[2]
    anak1[3] = ind2[3]
    anak2[0] = ind2[0]
    anak2[1] = ind2[1]
    anak2[2] = ind1[2]
    anak2[3] = ind1[3]
    return anak1, anak2
#Function mutasi
def mutasi(ind=[]):
    ind[1] = 1 - ind[1]
    return ind
#Function hitung Int
def hitungInt(ind = []):
    a = ind[3] * 1
    a = a + ind[2] * 2
    a = a + ind[1] * 4
    a = a + ind[0] * 8
    return a
#Buat function untuk melakukan hitung fitness
def hitungFitness(ind = []):
    a = hitungInt(ind)
    y = a**2 #sphere
    #y = np.sum(np.abs(a)) + np.product(np.abs(a)) #schwefel 2.22
    y = np.sum(100*(a+1)-(a))**2 + (a-1)**2 #Generalize rosenbrock
    \# y = \text{np.sum}((a^{**2}) - (10^{*}\text{math.cos}(2^{*}\text{math.pi}^*a)) + 10) \#\text{rastrigins}
    return y
i = 1
epochs = 99
MaxFitness = [0 for i in range(epochs)]
angkaFitness = [0 for i in range(epochs)]
#nilai fitness induk
for j in range(N):
    if (hitungFitness(induk[j])>MaxFitness[0]):
        MaxFitness[0] = hitungFitness(induk[j])
```

#Melakukan perulangan sampai 'epochs' kali

```
for i in range(30):
 while i < epochs:
  print("Iterasi: ", i+1)
  #Reproduksi
  a = random.random()
  if (a<Pc):
     #Cross over
     anak[0], anak[1] = crossover(induk[0], induk[1])
     anak[2], anak[3] = crossover(induk[0], induk[2])
     anak[4], anak[5] = crossover(induk[1], induk[2])
  else:
     #Mutasi
     for j in range(N):
        induk [j] = mutasi(induk[j])
  # ----- end reproduksi
  #Elistism
  minFitness = 10
  idx = 0
  idxanak = 0
  #Cari minFitness (induk terburuk)
  for j in range(len(induk)):
     #print(minFitness, hitungFitness(induk[j]))
     if (hitungFitness(induk[j])<minFitness):</pre>
        minFitness = hitungFitness(induk[j])
        idx = j
  #cari maxFit (Anak terbaik)
  maxFit = -10
  for j in range(len(anak)):
     if (hitungFitness(anak[j])>maxFit):
        maxFit = hitungFitness(anak[j])
        idxanak = j
  #Bandingkan individual replacement
  if (minFitness < maxFit):</pre>
     induk[idx] = anak[idxanak]
  #dari populasi induk yang baru mana yang paling baik
  for j in range(N):
     if (hitungFitness(induk[j])>MaxFitness[i]):
        MaxFitness[i] = hitungFitness(induk[j])
  i+=1
  print(MaxFitness)
   Iterasi: 37
   Iterasi: 38
   Iterasi: 39
   Iterasi: 40
   Iterasi: 41
   Iterasi: 42
```

```
Iterasi:
Iterasi: 44
Iterasi: 45
Iterasi: 46
Iterasi: 47
Iterasi: 48
Iterasi: 49
Iterasi: 50
Iterasi: 51
Iterasi: 52
Iterasi: 53
Iterasi: 54
Iterasi: 55
Iterasi: 56
Iterasi: 57
Iterasi: 58
Iterasi: 59
Iterasi: 60
Iterasi: 61
Iterasi: 62
Iterasi: 63
Iterasi: 64
```

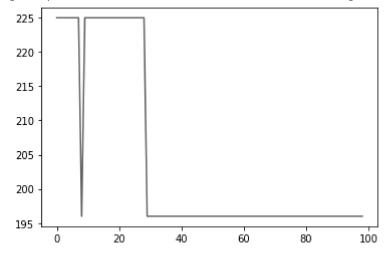
```
print("Nilai Mean : ", st.mean(MaxFitness))
print("Nilai Standard Deviation : ", st.stdev(MaxFitness))
print("Nilai Minimal : ",np.min(MaxFitness))
```

Nilai Mean : 204.2020202020202 Nilai Standard Deviation: 13.127304305382667

Nilai Minimal: 196

```
x = [i for i in range(epochs)]
plt.plot(x, MaxFitness)
```

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