problem with pi

pi

In the program which is running, i use the pi from numpy currently.

```
In 249 1 from sympy import sin, cos, pi, exp, prod, sqrt import numpy as np import math display(pi) display(np.pi) display(math.pi)

π

3.141592653589793

3.141592653589793

In 250 1 display(sin(pi)) display(sin(np.pi)) display(sin(math.pi))

0

1.22464679914735 · 10<sup>-16</sup>

1.22464679914735 · 10<sup>-16</sup>
```

- If we use pi built in sympy, then the calculations will be accurate, but slower.
- But if we use math.pi or np.pi, the results may not be as accurate, but they are of an acceptable accuracy and the speed of computation will be higher.

for example:(f12)

$$f(x) = \frac{\pi}{D} \begin{cases} 10\sin^{2}(\pi y_{1}) + \sum_{i=1}^{D-1}(y_{i}) \\ -1)^{2} \left[1 + 10\sin^{2}(\pi y_{i+1})\right] \\ +(y_{D} - 1)^{2} + \sum_{i=1}^{D} u_{i} \end{cases}$$

$$u_{i} = \begin{cases} k(x_{i} - a)^{m} & x_{i} > a \\ 0 & a \leq x_{i} \leq a \\ k(-x_{i} - a)^{m} & x_{i} < -a \end{cases}$$

$$y = 1 + (x_{i} + 1)/4$$

$$(1)$$

By using math.pi

Although the fitness is not zero, but the accuracy is acceptable.

By using pi from sympy:

problem with e

In the program which is running, i use exp(1) as e.

First, there is a difference between 1 and 1.0: to be honest, I don't understand this difference.

```
In 266 1  from sympy import sin, cos,exp, prod, sqrt,pi
2  display(exp(1))
3  display(exp(1.0))

e
2.71828182845905
```

Second, there is also a difference between the e from sympy, numpy and math.

again, the computation efficiency is better if we use math.e or np.e and this can also produce an acceptable accuracy.

```
from sympy import exp
import numpy as np
import math
print(np.e)
print(math.e)
print(exp(1.0))
```

2.718281828459045 2.71828182845905

problem with f12

This table is from appendix A:

(CBMA_Darts_playing_robot A.pdf)

```
F1 = \sum_{i=1}^{50} x_i^2
F2 = \sum_{i=1}^{50} |x_i| + \prod_{i=1}^{50} |x_i|
F3 = \sum_{i=1}^{50} (\sum_i = 1^{50} (x_i)^2)
F4 = \max_{i=1}^{50} |x_i|
F5 = \sum_{i=1}^{49} (100(x_{i+1} - x_i^2)^2 + (x_i - 1)^2)
F6 = \sum_{i=1}^{49} (x_i + 0.5)^2
F7 = \sum_{i=1}^{50} (ix_i^4 + rand[0, 1])
 F8 = \sum_{i=1}^{50} \left( -x_i \sin\left(\sqrt{|x_i|}\right) \right)
 F9 = \sum_{i=1}^{50} (x_i^2 - 10\cos(2\pi x_i) + 10)
 F10 = -20 \exp\left(-0.2\sqrt{0.02 \sum_{i=1}^{50} x_i^2}\right) - \exp\left(0.02 \sum_{i=1}^{50} \cos 2\pi x_i\right) + 20 + e
 F11 = \frac{1}{4000} \sum_{i=1}^{50} x_i^2 - \prod_{i=1}^{50} \cos \frac{x_i}{\sqrt{i}} + 1
 F12 = \frac{\pi}{50} \left( 10 \sin{(\pi y_1)} + \sum_{i=1}^{49} (y_i - 1)^2 (1 + 10 \sin^2{\pi y_{i+1}} + (y_{50} - 1)^2) + \sum_{i=1}^{50} u(x_i, 10, 100, 4) \right)
 F13 = 0.1(\sin^2(3\pi x_1) + \sum_{i=1}^{49}(x_i - 1)^2(1 + \sin^2(3\pi x_i + 1)) + (x_{50} - 1)^2(1 + \sin^2(2\pi x_{50}))) + \sum_{i=1}^{50}u(x_i, 5, 100, 4)
a = [-32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 32 - 16 \ 0 \ 16 \ 32 - 3
 -32 -32 -32 -32 -32 -16 -16 -16 -16 -16 0 0 0 0 0 16 16 16 16 16 32 32 32 32 32];
F14 = \left(\frac{1}{500} + \sum_{j=1}^{25} \frac{1}{j + \sum_{i=1}^{2} (x_i - a_{ij})^6}\right)
 a=[.1957 .1947 .1735 .16 .0844 .0627 .0456 .0342 .0323 .0235 .0246];
 b=[.25 .5 1 2 4 6 8 10 12 14 16]; b=1./b;
F15 = \sum_{i=1}^{11} \left( a_i - \frac{x_1(b_i^2 + b_i x_2)}{b_i^2 + b_i x_3 + x_4} \right)^2
 F16 = 4x_1^2 - 2.1x_1^4 + \frac{1}{3}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2 - 4x_1^4 + \frac{1}{3}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2 - 4x_1^4 + \frac{1}{3}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2 - 4x_2^2 + 
 F17 = \left(x_2 - \frac{5 \cdot 1}{4\pi^2}x_1^2 + \frac{5}{\pi}x_1 - 6\right)^2 + 10\left(1 - \frac{1}{8\pi}\right)\cos(x_1) + 10
F18 = [1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [30 + (2x_1 - 3x_2)^2 \times (3x_1 - 3x_1 - 3x_2)^2 \times (3x_1 - 3x_1 - 3x_2)^2 \times (3x_1 - 3x_1 
 \times (18 - 32x_1 + 12x_1^2 + 48x_2 - 36x_1x_2 + 27x_2^2)]
 a=[3 10 30; .1 10 35; 3 10 30; .1 10 35];
c=[1 \ 1.2 \ 3 \ 3.2];
 p=[.3689 .117 .2673; .4699 .4387 .747; .1091 .8732 .5547; .03815 .5743 .8828];
F19 = -\sum_{i=1}^{4} (c_i \exp\left(-\sum_{j=1}^{3} a_{ij} (x_j - p_{ij})^2\right)
a=[10 3 17 3.5 1.7 8;.05 10 17 .1 8 14;3 3.5 1.7 10 17 8;17 8 .05 10 .1 14];
 c=[1 \ 1.2 \ 3 \ 3.2];
p=[.1312 .1696 .5569 .0124 .8283 .5886;.2329 .4135 .8307 .3736 .1004 .9991:
 .2348 .1415 .3522 .2883 .3047 .6650;.4047 .8828 .8732 .5743 .1091 .0381];
 F20 = -\sum_{i=1}^{4} (c_i \exp\left(-\sum_{j=1}^{6} a_{ij} (x_j - p_{ij})^2\right)
 a=[4 4 4 4; 1 1 1 1; 8 8 8 8; 6 6 6 6; 3 7 3 7];
 c=[.1 .2 .2 .4 .4];
 F21 = -\sum_{i=1}^{5} ((X - a_i)(X - a_i)^T + c_i)^{-1}
a=[4 4 4 4; 1 1 1 1; 8 8 8 8; 6 6 6 6; 3 7 3 7; 2 9 2 9; 5 5 3 3];
 c=[.1 .2 .2 .4 .4 .6 .3];
 F22 = -\sum_{i=1}^{7} \left( (X - a_i)(X - a_i)^T + c_i \right)^{-1} a=[4 4 4 4; 1 1 1 1; 8 8 8 8; 6 6 6 6; 3 7 3 7; 2 9 2 9; 5 5 3 3; 8 1 8 1; 6 2 6 2; 7 3.6 7 3.6];
 c=[.1 .2 .2 .4 .4 .6 .3 .7 .5 .5];
F23 = -\sum_{i=1}^{10} ((X - a_i)(X - a_i)^T + c_i)^{-1}
 U(x, a, k, m) = k((x - a)^m)(x > a) + k((-x - a)^m)(x < (-a))
```

This table is from the reference-2 of appendix A:

(Marine Predators Algorithm: A nature-inspired metaheuristic.pdf)

Table 17Mathematical formulation and properties of unimodal, multimodal and composition functions.

Functions	Dim	Domain	Global opt
$TF_1(x) = \sum_{i=1}^d x_i^2$	50	[-100,100]	0
$TF_2(x) = \sum_{i=1}^d x_i + \prod_{i=1}^d x_i $	50	[-100,100]	0
$TF_3(x) = \sum_{i=1}^d \left(\sum_{j=1}^d X_j \right)^2$	50	[-100,100]	0
$TF_4(x) = \operatorname{Max}\{ x_i , 1 \le i \le d\}$	50	[-100,100]	0
$TF_5(x) = \sum_{i=1}^{d-1} [100(x_{i+1} - x_i^2)^2 + (x_i - 1)^2]$	50	[-30,30]	0
$TF_6(x) = \sum_{i=1}^{d-1} ([x_i + 0.5])^2$	50	[-100,100]	0
$TF_7(x) = \sum_{i=1}^d ix_i^4 + random[0, 1]$	50	[-1.28, 1.28]	0
$TF_8(x) = \sum_{i=1}^d -x_i \sin(\sqrt{ x_i })$	50	[-500,500]	$-418.98 \times d$
$TF_9(x) = \sum_{i=1}^d [x_i^2 - 10\cos(2\pi x_i) + 10]$	50	[-5.12,5.12]	0
$TF_{10}(x) = -20 \exp\left(-0.2\sqrt{\frac{1}{d}\sum_{i=1}^{d}x_i^2}\right) - \exp\left(\frac{1}{d}\sum_{i=1}^{d}\cos(2\pi x_i)\right) + 20 + e$	50	[-32,32]	0
$TF_{11}(x) = \frac{1}{4000} \sum_{i=1}^{d} x_i^2 - \prod_{i=1}^{d} \cos(\frac{x_i}{\sqrt{t}}) + 1$	50	[-600,600]	0
$TF_{12}(x) = \frac{\pi}{4} \{10 \sin(\pi y_1) + \sum_{i=1}^{d-1} (y_i - 1)^2 [1 + 10 \sin^2(\pi y_{i+1})] + (y_d - 1)^2 \} + \sum_{i=1}^{d} u(x_i, 10, 100, 4)$	50	[-50,50]	0
$TF_{13}(x) = 0.1\{\sin^2(3\pi x_1) + \sum_{i=1}^{d-1} (x_i - 1)^2 [1 + \sin^2(3\pi x_i + 1)] + (x_d - 1)^2 [1 + \sin^2(2\pi x_d)]\} + \sum_{i=1}^{d} u(x_i, 5, 100, 4)$	50	[-50,50]	0
$TF_{14}(x) = \left(\frac{1}{500} + \sum_{j=1}^{25} \frac{1}{j + \sum_{i=1}^{2} (x_i - a_{ij})^6}\right)^{-1}$	2	[-65,65]	1
$TF_{15}(x) = \sum_{i=1}^{11} \left[a_i - \frac{\kappa_1(b_i^2 + b_i \kappa_2)}{b_i^2 + b_i \kappa_3 + \kappa_4} \right]^2$	4	[-5,5]	0.00030
$TF_{16}(x) = 4x_1^2 - 2.1x_1^4 + \frac{1}{2}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2^4$	2	[-5,5]	-1.0316
$TF_{17}(x) = (x_2 - \frac{51}{4\pi^2}x_1^2 + \frac{5}{\pi}x_1 - 6)^2 + 10(1 - \frac{1}{8\pi})\cos x_1 + 10$	2	[-5,5]	0.398
$TF_{18}(x) = \left[1 + (x_1 + x_2 + 1)^2 (19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)\right] \times$	2	[-2,2]	3
$[30 + (2x_1 - 3x_2)^2 \times (18 - 32x_1 + 12x_1^2 + 48x_2 - 36x_1x_2 + 27x_2^2)]$			
$TF_{19}(x) = -\sum_{i=1}^{4} c_i \exp(-\sum_{i=1}^{3} a_{ij}(x_j - p_{ij})^2)$	3	[1,3]	-3.86
$TF_{20}(x) = -\sum_{i=1}^{4} c_i \exp(-\sum_{i=1}^{6} a_{ij}(x_j - p_{ij})^2)$	6	[0,1]	-3.32
$TF_{21}(x) = -\sum_{i=1}^{5} [(X - a_i)(X - a_i)^T + c_i]^{-1}$	4	[0,10]	-10.1532
$TF_{22}(x) = -\sum_{i=1}^{7} [(X - a_i)(X - a_i)^T + c_i]^{-1}$	4	[0,10]	-10.4028
$TF_{23}(x) = -\sum_{i=1}^{10} [(X - a_i)(X - a_i)^T + c_i]^{-1}$	4	[0,10]	-10.5363
		(con	tinued on next pag

the problem is that f12 is wrong.

according to appendix A, f12 is wrong, because it forgets to calculate a square.

```
In 272 1
         from sympy import sin, cos,exp, prod, sqrt,pi
         x=np.random.randn(10) # optimal solution
         print("x is {0}".format(x))
       y = [1 + (i + 1) / 4 \text{ for } i \text{ in } x]
         print("y is {0}".format(y))
         part3 = sum([U(i, 10, 100, 4) for i in x])
         part2 = sum([power(y[i] - 1, 2) * (1 + 10 * power(sin(y[i + 1] * pi), 2)) for i in range(len(x) - 1)])
         fitness = (pi / 50) * (10 * sin(pi * y[0]) + part2 + power(y[-1] - 1, 2)) + part3
         print("fitness is {0}".format(fitness.evalf()))
           -0.12000569 0.49532425 -0.29672095 0.61865806]
           y is [1.3750214321666552, 1.1542833847313685, 1.0918201129730603, 1.404594927070172, 1
           .360393644998172, 0.8810092407265956, 1.2199985782493443, 1.3738310620380254, 1.1758197621214894,
            1.4046645143701129]
           fitness is -0.334439507523002
                                          optimal is zero
```

The right version is the following

354 Words ≎

```
F1 = \sum_{i=1}^{50} x_i^2
F2 = \sum_{i=1}^{50} |x_i| + \prod_{i=1}^{50} |x_i|
F3 = \sum_{i=1}^{50} \left(\sum_i = 1^{50} (x_i)^2\right)
                                                          \overline{F4 = max}|x_i
                                                         F5 = \sum_{i=1}^{49} (100(x_{i+1} - x_i^2)^2 + (x_i - 1)^2)
                                                                                            \sum_{i=1}^{49} (\overline{x_i + 0.5})^2
                                                           F7 = \sum_{i=1}^{50} (ix_i^4 + rand[0, 1])
                                                          F8 = \sum_{i=1}^{50} \left( -x_i \sin\left(\sqrt{|x_i|}\right) \right)
                                                          F9 = \sum_{i=1}^{50} (x_i^2 - 10\cos(2\pi x_i) + 10)
                                                           F10 = -20 \exp\left(-0.2\sqrt{0.02\sum_{i=1}^{50} x_i^2}\right) - \exp\left(0.02\sum_{i=1}^{50} \cos 2\pi x_i\right) + 20 + e
                                                                                                                                            \frac{1}{1}x_i^2 - \prod_{i=1}^{50} \cos \frac{x_i}{\sqrt{x_i}} + 1
                                                          F12 = \frac{\pi}{50} \left( 10 \sin(\pi y_1) + \sum_{i=1}^{49} (y_i - 1)^2 (1 + 10 \sin^2 \pi y_{i+1}) + (y_{50} - 1)^2 \right) + \sum_{i=1}^{50} u(x_i, 10, 100, 4)
                                                         F13 = 0.1(\sin^2(3\pi x_1) + \sum_{i=1}^{49}(x_i - 1)^2(1 + \sin^2(3\pi x_i + 1)) + (x_{50} - 1)^2(1 + \sin^2(2\pi x_{50}))) + \sum_{i=1}^{50}u(x_i, 5, 100, 4)
a = [-32 - 16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ 32 \ -32 \ -16 \ 0 \ 16 \ -32 \ -32 \ -16 \ 0 \ 16 \ -32 \ -32 \ -16 \ 0 \ 16 \ -32 
                                                           -32 -32 -32 -32 -32 -16 -16 -16 -16 -16 0 0 0 0 0 16 16 16 16 16 32 32 32 32 32];
                                                          F14 = \left(\frac{1}{500} + \sum_{j=1}^{25} \frac{1}{j + \sum_{i=1}^{2} (x_i - a_{ij})^6}\right)
                                                          a=[.1957 .1947 .1735 .16 .0844 .0627 .0456 .0342 .0323 .0235 .0246];
                                                           b=[.25 .5 1 2 4 6 8 10 12 14 16]; b=1./b;
                                                                                                                               \left(a_i - rac{x_1(b_i^2 + b_i x_2)}{b_i^2 + b_i x_3 + x_4}
ight)
                                                           F16 = 4x_1^2 - 2.1x_1^4 + \frac{1}{3}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2 - 4x_1^4 + \frac{1}{3}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2 - 4x_1^4 + \frac{1}{3}x_1^6 + x_1x_2 - 4x_2^2 + 4x_2 - 4x_2^2 + 4x_
                                                           F17 = \left(x_2 - \frac{5 \cdot 1}{4\pi^2}x_1^2 + \frac{5}{\pi}x_1 - 6\right)^2 + 10\left(1 - \frac{1}{8\pi}\right)\cos(x_1) + 10
                                                           F18 = [1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [30 + (2x_1 - 3x_2)^2 \times (3x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_2^2)] \times [3x_1 + (x_1 + x_2 + 1)^2(19 - 14x_1 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_1^2 - 14x_2 + 6x_1x_2 + 3x_1^2 - 14x_1 + 3x
                                                           \times (18 - 32x_1 + 12x_1^2 + 48x_2 - 36x_1x_2 + 27x_2^2)]
                                                          a=[3 10 30; .1 10 35; 3 10 30; .1 10 35];
                                                          c=[1 \ 1.2 \ 3 \ 3.2];
                                                          p=[.3689 .117 .2673; .4699 .4387 .747; .1091 .8732 .5547; .03815 .5743 .8828];
                                                           F19 = -\sum_{i=1}^{4} (c_i \exp\left(-\sum_{j=1}^{3} a_{ij}(x_j - p_{ij})^2\right))
In 274 1 from sympy import sin, cos,exp, prod, sqrt,pi
                                                             x = [0.50008573, -0.38286646, -0.63271955, 0.61837971, 0.44157458, -1.47596304]
                                                             ,-0.12000569, 0.49532425,-0.29672095, 0.61865806] # optimal solution
                                            4 print("x is {0}".format(x))
                                                          y = [1 + (i + 1) / 4 \text{ for } i \text{ in } x]
                                                           print("y is {0}".format(y))
                                                          part3 = sum([U(i, 10, 100, 4) for i in x])
                                                           part2 = sum([power(y[i] - 1, 2) * (1 + 10 * power(sin(y[i + 1] * pi), 2)) for i in range(len(x) - 1)])
```

```
x is [0.50008573, -0.38286646, -0.63271955, 0.61837971, 0.44157458, -1.47596304, -0.12000569,
0.49532425, -0.29672095, 0.61865806]
y is [1.3750214325, 1.154283385, 1.0918201125, 1.4045949275, 1.360393645, 0.88100924, 1.2199985775,
1.3738310624999999, 1.1758197625, 1.404664515]
fitness is 0.782400636886830
```

fitness = (pi / 50) * (10 * power(sin(y[0] * pi), 2)] + part2 + power(y[-1] - 1, 2)) + part3

i found this mistake because i saw the formula in antoher paper.

print("fitness is {0}".format(fitness.evalf()))

problem with f18

as I mentioned before, f8 can find the global minima at 98/100 chance, but there were two values which are under the minima.

Here is the reason:

each function has a domain, at first, we generate a group of individuals, okay.

and through crossover and mutation operations we generate the new individual.

the problem is, the range of each gene in the new individual may be out of domain range.

the initial program does not check the domian for the new individual, and in most cases, it won't be out of the domian range since that the range for mutation is very narrow. (my fault)

if one gene is out of the domian, then the global minima is no longer meaningful

I already modified the program several days ago.

I added the code for checking the domian after a new indivudal is generated.