

PSO

May 22, 2021

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[1]: from random import random
      from random import uniform
      from numpy.random import normal
      import math
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[2]: # functions to optimize (minimize)
      def square(x):
          total=0
          for i in range(len(x)):
              total+=x[i]**2
          return total

      def rosenbrock(x):
          a = 1
          b = 15
          return ((a - x[0]**2)+b*((x[1]-x[0]**2)**2))
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[3]: class Particle:
      def __init__(self, initial_pos):
          self.position_i=[]          # particle position
          self.velocity_i=[]          # particle velocity
          self.pos_best_i=[]          # best position individual
          self.err_best_i=-1          # best error individual
          self.err_i=-1               # error individual

          for i in range(0,num_dimensions):
              self.velocity_i.append(float(normal(0.5,0.175,1)))
              self.position_i.append(initial_pos[i])

      def evaluate(self,cost_function):
          '''
          evaluate current fitness

          :params
          cost_function : function to optimize
          '''
          self.err_i=cost_function(self.position_i)
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        # check to see if the current position is an individual best
        if self.err_i < self.err_best_i or self.err_best_i == -1:
            self.pos_best_i = self.position_i.copy()
            self.err_best_i = self.err_i

    def update_velocity(self, pos_best_g):
        """
        update new particle velocity

        :params
        pos_best_g : global best position
        """
        w = uniform(0.4, 0.9)          # linearly varied b/w 0.9 to 0.4
        c1 = 2
        c2 = 2

        for i in range(0, num_dimensions):
            r1 = random()
            r2 = random()

            vel_cognitive = c1 * r1 * (self.pos_best_i[i] - self.position_i[i])
            vel_social = c2 * r2 * (pos_best_g[i] - self.position_i[i])
            self.velocity_i[i] = w * self.velocity_i[i] + vel_cognitive + vel_social

    def update_position(self, bounds):
        """
        updates the particle position based on new velocity updates

        :params
        bounds
        """
        for i in range(0, num_dimensions):
            self.position_i[i] = self.position_i[i] + self.velocity_i[i]

            # check boundary conditions
            if self.position_i[i] > bounds[i][1]:
                self.position_i[i] = bounds[i][1]
            if self.position_i[i] < bounds[i][0]:
                self.position_i[i] = bounds[i][0]

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[4]: def minimize(cost_function, initial_pos, bounds, num_particles, max_iterations,
    verbose=False):
    global num_dimensions

    num_dimensions = len(initial_pos)
    err_best_g = -1          # best error for group

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pos_best_g=[]                                # best position for group

# create the swarm
swarm=[]
for i in range(0,num_particles):
    swarm.append(Particle(initial_pos))

i=0
while i<max_iterations:
    if verbose: print(f'iteration: {i:>4d}, best solution: {err_best_g:10.
↪6f}')

    # evaluate fitness
    for j in range(0,num_particles):
        swarm[j].evaluate(cost_function)

        # determine if current particle is the best (globally)
        if swarm[j].err_i<err_best_g or err_best_g==-1:
            pos_best_g=list(swarm[j].position_i)
            err_best_g=float(swarm[j].err_i)

    # update velocities and position
    for j in range(0,num_particles):
        swarm[j].update_velocity(pos_best_g)
        swarm[j].update_position(bounds)
    i+=1

return err_best_g, pos_best_g

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[5]: initial=[5,5]                            # initial starting location [x1,x2...]
    bounds=[(-10,10),(-10,10)] # input bounds [(x1_min,x1_max),(x2_min,x2_max)...]

    # for rosenbrock function
    minima, best_position = minimize(rosenbrock, initial, bounds, num_particles=15,
↪max_iterations=30, verbose=True)
    print('\n\nBest Position:',best_position)
    print('Best Solution:',minima)

```

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iteration:    0, best solution:  -1.000000
iteration:    1, best solution: 5976.000000
iteration:    2, best solution: 5976.000000
iteration:    3, best solution: 1951.321343
iteration:    4, best solution:  61.556599
iteration:    5, best solution:   3.752611
iteration:    6, best solution:   3.752611
iteration:    7, best solution: -3.899044
iteration:    8, best solution: -3.920926

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iteration:    9, best solution:  -3.920926
iteration:   10, best solution:  -3.920926
iteration:   11, best solution:  -4.964809
iteration:   12, best solution:  -5.064727
iteration:   13, best solution:  -5.064727
iteration:   14, best solution:  -5.064727
iteration:   15, best solution:  -5.064727
iteration:   16, best solution:  -5.064727
iteration:   17, best solution:  -5.064727
iteration:   18, best solution:  -5.274907
iteration:   19, best solution:  -5.274907
iteration:   20, best solution:  -5.424370
iteration:   21, best solution:  -5.608808
iteration:   22, best solution:  -5.608808
iteration:   23, best solution:  -5.608808
iteration:   24, best solution:  -5.608808
iteration:   25, best solution:  -5.608808
iteration:   26, best solution:  -5.608808
iteration:   27, best solution:  -5.608808
iteration:   28, best solution:  -5.608808
iteration:   29, best solution:  -5.608808

```

Best Position: [2.586473862228981, 6.61634458572222]
Best Solution: -5.6088078782347335

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[6]: # for square function
minima_sq, best_position_sq = minimize(square, initial, bounds,
    ↳ num_particles=15, max_iterations=30, verbose=True)
print('\n\nBest Position:', best_position_sq)
print('Best Solution:', minima_sq)

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iteration:    0, best solution:  -1.000000
iteration:    1, best solution:  50.000000
iteration:    2, best solution:  50.000000
iteration:    3, best solution:  40.451234
iteration:    4, best solution:  26.225596
iteration:    5, best solution:  13.825216
iteration:    6, best solution:   5.020039
iteration:    7, best solution:   1.778763
iteration:    8, best solution:   1.318052
iteration:    9, best solution:   0.652904
iteration:   10, best solution:   0.416916
iteration:   11, best solution:   0.129198
iteration:   12, best solution:   0.008798
iteration:   13, best solution:   0.008336
iteration:   14, best solution:   0.008336
iteration:   15, best solution:   0.006516

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iteration: 16, best solution: 0.002443
iteration: 17, best solution: 0.002443
iteration: 18, best solution: 0.001535
iteration: 19, best solution: 0.001535
iteration: 20, best solution: 0.001535
iteration: 21, best solution: 0.001535
iteration: 22, best solution: 0.000919
iteration: 23, best solution: 0.000919
iteration: 24, best solution: 0.000919
iteration: 25, best solution: 0.000919
iteration: 26, best solution: 0.000919
iteration: 27, best solution: 0.000919
iteration: 28, best solution: 0.000102
iteration: 29, best solution: 0.000035
```

Best Position: [-0.003899158934036445, 0.004423475758581159]

Best Solution: 3.477057817963139e-05

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