Particle Swarm Optimization (PSO)

```
In [1]: import numpy as np
    from numpy import absolute
    from numpy.random import uniform, choice
    from random import randint

import seaborn as sns
    import matplotlib.pyplot as plt

from IPython.display import HTML, display
    from pprint import pprint
    from tabulate import tabulate

In [2]: # Codes Implemented
    from python_codes.evaluation import fitness_fx
    from python_codes.particle import Particle

In [3]: # Algorithm parameters
    (b_lo, b_up) = (-1, 2) # Lower and up boundaries
    n_dimensions = 5 # Number of genes per particle
    n_particles = 100 # Number of particles in population
    swam_size = 2 # Number of neighbors for each particle (swarm)
    n_iters = 500 # Number of iterations (criterion)
    g = None # Best known position (vector)
    omega = 1 # Omega contant
    phi_p = 0.5 # Phi p constant
    phi_g = 0.5 # Phi g constant
```

Next we generate the initial population:

```
In [4]: # Generate particle population
        particle_pop = []
        for i in range(n particles):
            part velocity = uniform(-absolute(b lo - b up), absolute(b lo - b up), size=r
            part_position = uniform(low=b_lo, high=b_up, size=n_dimensions)
            best_position = part_position
            if g is not None:
                if fitness_fx(best_position).sum() > fitness_fx(g).sum():
                    g = best_position.copy()
            else:
                g = best_position.copy()
            p = Particle(position=part position, velocity=part velocity, best position=be
            particle_pop.append(p)
        # Define the swarm of each particle
        for particle in particle_pop:
            neighbors = choice(particle_pop, size=swam_size)
            particle.neighbors particles = neighbors
```

```
In [10]: print('Best known position (g) in the initial population:', g)
print('f(g): ', fitness_fx(g).sum())
```

```
Best known position (g) in the initial population: [1.66226588 1.63457626 1.825 89829 1.87030108 0.86681045] f(g): 11.565157057023557
```

Now that we've a initial population, the algorithm its execute. That algorithm its based on the pseudocode, shown in the beginning of the notebook:

```
In [6]: best per it = []
        for it in range(n iters):
             for particle in particle pop:
                 # Get the current values
                 tmp_vel = particle.velocity.copy()
                 tmp position = particle.position.copy()
                 tmp best pos = particle.best position.copy()
                 # Update particle velociy
                 new velocity = []
                 for d in range(n_dimensions):
                     r_p, r_g = uniform(), uniform()
                     v_id = omega * tmp_vel[d] + phi_p * r_p * (
                         tmp_best_pos[d] - tmp_position[d]) + phi_g * r_g * (g[d] - tmp_position[d])
                     new_velocity.append(v_id)
                 particle.velocity = np.array(new_velocity)
                 # Update particle position
                 tmp position += new velocity
                 # If any dimension overcome the limits,
                 # other values in the limits its generate for this dimension.
                 tmp position = [x \text{ if } (-1 \le x \le 2) \text{ else uniform(low=b lo, high=b up) for}]
                 particle.position = np.array(tmp_position)
                 # Update best positions
                 if fitness fx(particle.position).sum() > fitness fx(particle.best position
                     # Update the particles best known position
                     particle.best position = particle.position
                     if fitness_fx(particle.best_position).sum() > fitness_fx(g).sum():
                         # Update the swarms best known position
                         g = particle.best position
            best per it.append(g)
```

```
In [11]: print('Best known position (g) after 500 iterations:', g)
print('f(g): ', fitness_fx(g).sum())
```

```
Best known position (g) after 500 iterations: [1.66226588 1.63457626 1.82589829 1.87030108 0.86681045] f(g): 11.565157057023557
```

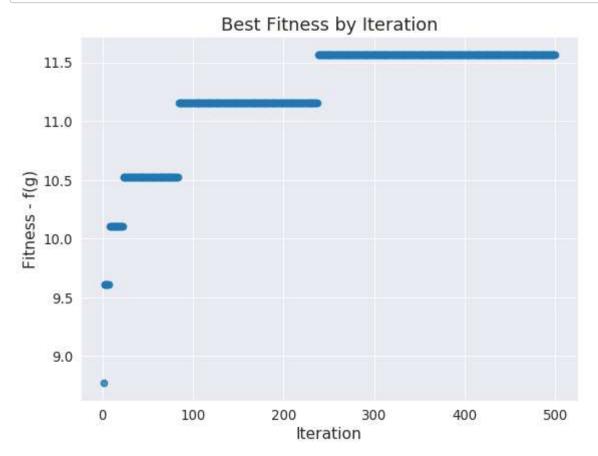
Next, we show the results by iteration. We can notice that the g position merge to a local minima.

```
In [8]: sns.set_style('darkgrid')

plt.figure(num=None, figsize=(8, 6), dpi=80, facecolor='w', edgecolor='k')

y = np.array([fitness_fx(x).sum() for x in best_per_it])
x = np.array(range(1, len(y)+1))

sns.regplot(x=x, y=y, fit_reg=False)
plt.title('Best Fitness by Iteration', fontsize=16)
plt.xlabel('Iteration', fontsize=14)
plt.ylabel('Fitness - f(g)', fontsize=14)
plt.yticks(fontsize=12)
plt.xticks(fontsize=12)
plt.show()
```



WIP

```
In []: # WIP: function to show points moving in each iteration...
%matplotlib inline
import time
import pylab as pl
from IPython import display

for i in range(10):
    pl.plot(pl.randn(100))
    display.clear_output(wait=True)
    display.display(pl.gcf())
    time.sleep(1.0)
```