## Homework 6

## Particle Swarm Optimization using Schwefel Benchmark Function

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## ${\rm April}\ 2022$

## Contents

L	Que	estion 1: Particle Swarm Optimization	2
	1.1	Schwefel Function	3
	1.2	Functions for the Global and Local Best Logic (using Ring Structure)	3
	1.3	Function to Initialize the Swarm	5
	1.4	Function to Update Velocity and Position	6
	1.5	Function to Calculate Particle Historical Best Fitness Value and Position $\dots \dots \dots$	8
	1.6	2 Functions to Display Best Solution and Write to CSV $\ \ldots \ \ldots \ \ldots \ \ldots \ \ldots$	9
	1.7	Driver Funtion to Compile and Loop through a PSO Problem	11
	1.8	Part 1 (b) & (c) - Global Best Method	13
		1.8.1 1(b) 2D Global Swarm	13
		1.8.2 1(b) i iii. Iteration Table & Plot	14
		1.8.3     1(c) 200D Global Swarm w/Various Param    .    .	17
	1.9	Part 1 (d) & (e) - Local Best Method $\  \   \dots $	19
		1.9.1 Reference: Local Swarm Calculation and Logic	19
		1.9.2 1(d) 2D Local Swarm	20
		1.9.3 1(e) 200D Local Swarm w/Various Params	21

#### **Global Variables**

Input variables like the *random seed*, and the *lower and upper bound of solution space* Please assume these are referenced by following code chunks

```
import math
from random import Random
import numpy as np

# Random seed
seed = 12345
randNumGenerator = Random(seed)

# bounds for Schwefel Function search space
lowerBound = -500
upperBound = 500
```

## 1 Question 1: Particle Swarm Optimization

Find the global optimium of the Schwefel function below

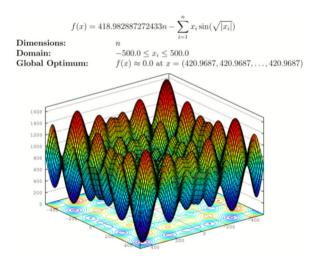


Figure 1: 2D Schwefel Function

#### 1.1 Schwefel Function

#### 1.2 Functions for the Global and Local Best Logic (using Ring Structure)

- Local best topology method: Ring structure
- The driver function takes in a string called local or global
- This parameter switches between these two methods for the driver
- For example, if I chose local, then it would use the local best method
- See inline comments for detailed logic

```
# GLOBAL MIN VALUE AND POSITION SEARCH FUNCTION
# Returns the 2 element list (each containing a single value) with the global best particle's:
# ---- [0] min value and
# ---- [1] associate position of
def getGlobalBest(fitnessValues, positions, swarmSize):
   minValue = np.min(fitnessValues)
                                      # Find the Minimum fitness value of all particles
   minIndex = fitnessValues.index(minValue) # Find the index of the position for the min. fit. value
   minPosition = positions[minIndex][:] # Now get a copy of the particle's position with min index
   # Returns: the global best particle's minimum fitness value and its position
   return [minValue, minPosition]
# LOCAL MIN VALUE AND POSITION SEARCH FUNCTION
# Topology: Ring structure with n neighbors (default 2)
# Returns the 2 element list of lists with the each particle's local best within neighborhood
# ---- [0] min value and
# ---- [1] associate position of
```

```
# Can change numParticlesInNbrhood to consider more or less in particle's neighborhood
# -----
def getLocalBest(fitnessValues, positions, swarmSize,
                numParticlesInNbrhood = 2): # Number of particles to compare to for local best
   lBestFitValue = [] # will hold the best VALUE of the n surrounding particles, for each particle
   lBestPosition = [] # will hold the best POSITION of the n surrounding particles, for each particle
   # For every particle in the swarm, (starting at n less than index 0)
   for particle in range(-numParticlesInNbrhood, swarmSize - numParticlesInNbrhood):
       # Identify the two neighbors fitness value of this particle,
       # which are the two preceding particles
       personalBestNeighbor1 = fitnessValues[particle]
       personalBestNeighbor2 = fitnessValues[particle + 1]
       # Identify the lowest fitness value of this particle's the two preciding neighbors
       minNeighValue = min(personalBestNeighbor1, personalBestNeighbor2)
       # Store the index of the particle
       minNeighIndex = fitnessValues.index(minNeighValue)
       # Store the particle's best neighbors fitness value and position
       lBestFitValue.append(fitnessValues[minNeighIndex])
       lBestPosition.append(positions[minNeighIndex])
   # Returns a list of particles and the min of their n best fit. valued neighbors
   return[lBestFitValue, lBestPosition]
# If you needed to index the list just returned for global or local best
VALUE IDX
POSITION_IDX = 1
```

#### 1.3 Function to Initialize the Swarm

• Allows you to easily change the swarm size, dimensions, an the local or global best method

```
# STEP 1 - SWARM INITIALIZATION / EVALUATION
# Randomly initialize a swarm instance
# Set the partical's best to it's starting position
# -----
def initializeSwarm(swarmSize, numDimensions, functionToGetBest):
   # In the current time period, position[particle] and velocity[particle] of each particle i,
   # Each particle contains n-dimensional list of the coordinate position & velocity
   position = [[] for _ in range(swarmSize)] # X[particle]: position (2D: x, y) of particle i
   velocity = [[] for _ in range(swarmSize)] # V[particle]: velocity (2D: x, y) of particle i
   # Lists containing info related to each particle in swarm
   pCurrFitValue = [] # X[particle] The current position of particle i
   # For each particle and dimension, randomly initialize the...
   for particle in range(swarmSize):
       for theDimension in range(numDimensions):
           # Position: give random value between lower/upper bounds (-500, 500 for schwefel)
           position[particle].append(randNumGenerator.uniform(lowerBound, upperBound))
           # Velocity: give random value between -1 and 1 --- maybe these are good bounds? maybe no
           velocity[particle].append(randNumGenerator.uniform(-1, 1))
       # 1.1 - Evaluate fitness value
       pCurrFitValue.append(evalFitnessVal(position[:][particle])) # evaluate the current position's
   # 1.2 - Log the individual and global bests
   pBestPosition = position[:] # initialize pBestPosition to the starting position
   pBestFitValue = pCurrFitValue[:] # initialize pBestPosition to the starting position's value
   # 1.3 - Log the Global or local best (depends on chosen method) fitness value and position
   glBestFitValue, glBestPosition = functionToGetBest(pBestFitValue[:], pBestPosition[:], swarmSize)
   return [position, velocity, pCurrFitValue,
          pBestPosition, pBestFitValue,
           glBestFitValue, glBestPosition]
```

#### 1.4 Function to Update Velocity and Position

- Uses the velocity calcultion to update the position.
- Function is dynamic and can use the local or global best methods
- Also handles:
  - Infeasibility of Velocity: pushes the position back into the lower or upper bound solution space (see logic below)
  - Infeasibility of Position: Randomly moves above the solution space until the position has found a feasible solution

```
# UPDATE VELOCITY AND POSITION
# -----
def updateVelocityAndPosition(intertiaWeight, velocity, position, phi1, phi2, pBestPosition, glBestPosi
                           swarmSize, numDimensions):
# Velocity -----
   ## random weights of r for random velocity adjustment
   r1, r2 = randNumGenerator.random(), randNumGenerator.random()
   ## Calculations of updating velocity, separated by
   ## intertia + cognitive + social (for simplicity)
   vInertia = np.multiply(intertiaWeight, velocity[:])
                                                                            # Interia
                                                                                       componen
   vCognitive = np.multiply(phi1*r1, np.subtract( pBestPosition[:], position[:])) # Cognitive componen
   vSocial = np.multiply(phi2*r2, np.subtract(glBestPosition[:], position[:])) # Social
                                                                                       componen
   ## Update the new velocity to the summation of intertia, cognitive, and social
   newVelocity = vInertia[:] + vCognitive[:] + vSocial[:]
   ## Limit the velocity between the upper and lower bound limits
   for particle in range(swarmSize):
       for theDimension in range(numDimensions):
           # If the new velocity of particle i is > the limit, then reduce to the limit
           if newVelocity[particle][theDimension] > upperBound:
              newVelocity[particle][theDimension] = upperBound
           # If the new velocity of particle i is < the limit, then increase to the limit
           if newVelocity[particle][theDimension] < lowerBound:</pre>
              newVelocity[particle][theDimension] = lowerBound
   # Position -----
   ## Update new position based on the updated velocity
   newPosition = position[:] + newVelocity[:]
   ## Make sure that the position is within the bounds ------
   for particle in range(swarmSize):
       for theDimension in range(numDimensions):
           # Check to see if position in bounds
           while (newPosition[particle][theDimension] > upperBound) or (newPosition[particle][theDimen
```

#### 1.5 Function to Calculate Particle Historical Best Fitness Value and Position

```
# Compare current position fitness value to the current best (for each particle)
# ------
def calculateParticleBests(position, swarmSize, numDimensions,
                         pCurrFitValue, pBestPosition, pBestFitValue):
   # Calculate the fitness of the new positions
   for particle in range(swarmSize):
       for theDimension in range(numDimensions):
           # Get the current fitness value of the new positions
           pCurrFitValue[particle] = evalFitnessVal(position[:][particle])
           # Compare the current positions' value to their person best
           if pCurrFitValue[particle] < pBestFitValue[particle]:</pre>
               # If better, then set the best VALUE to the current value (as a copy [:])
               pBestFitValue[particle] = pCurrFitValue[:][particle]
               # If better, then set the best POSITION to the current position (as a copy [:])
               pBestPosition[particle] = position[:][particle]
   return [pCurrFitValue, pBestPosition, pBestFitValue]
```

#### 1.6 2 Functions to Display Best Solution and Write to CSV

- writeIteratonsToCSV() is used to write a CSV with the top n iterations
- Using R, I plotted the first 5 iterations of swarmSize 5 (see related question below)

```
# ------
# DISPLAY GLOBAL BEST AND DIMENSIONS FUNCTION
# Function for displaying the global best and its dimensions
# -----
def displayGlobalBest(glBestFitValue, glBestPosition, numDimensions, printDims):
   # Print the global optima
   print('\nGlobal Best Value:\t % 0.4f' % glBestFitValue)
   # Print each dimension (if toggled)
   if printDims:
      print('For each [dimension], Global Best Position:')
       # Print the position of each dimension in markdown table format
      print('\n',
            '| Dimension | Global Best | ', sep = '')
      print('|-----|')
      for theDimension in range(numDimensions):
          print('|' + str(theDimension).rjust(10, ' '),
               '|' + '{:.4f}'.format(glBestPosition[theDimension]).rjust(12, ' ') + ' |'
# WRITE TOP n SWARM ITERATIONS TO A CSV FUNCTION
# Basic function for writing to file
# -----
def writeIteratonsToCSV(numDimensions, # Number of dimensions in the swarm
                    filename,
                    method,
                    velocityIterations,
                    positionIterations,
                    gBestPositionIterations,
                    swarmSize):
   iterBreak
             = 1 # Display every n iterations
   maxIterToView = 5 # Top iteration to display
   # Write to file if the dimensions are 2D and using Global best.
   # (only global since list structure for each particle, not single value)
   if numDimensions == 2 and method == 'global':
       # Print the first 5 best positions of the swarm, while highlighting global best
      f = open(filename + '.csv', 'w') # Open CSV for writing
       # Column names
      f.write('iteration,particle,position1,position2,globalBest1,globalBest2,velocity1,velocity2\n')
       # Write the data to a CSV for plotting and summary table
```

```
for iteration in range(maxIterToView):
    thisIter = ''
    for particle in range(swarmSize):
        if iteration % iterBreak == 0:
            # Get the Iteration, Particle, position1/2 and velocity 1/2
            theVelocities = velocityIterations[iteration][particle]
            thePositions = positionIterations[iteration][particle]
            globalBest
                       = gBestPositionIterations[iteration]
            # Convert the positions to flat text
                      = ','.join(['{:.4f}'.format(positionVal) for positionVal in theVelocitie
                      = ','.join(['{:.4f}'.format(positionVal) for positionVal in thePositions
            posDims
            globalDims = ','.join(['{:.4f}'.format(positionVal) for positionVal in globalBest])
            # Write tp the file
            f.write(str(iteration+1) + ',' + str(particle) + ',' + str(posDims) + ',' +
                    str(globalDims) + ',' + str(velDims) + ' n')
f.close() # close the file for saving output
```

#### 1.7 Driver Funtion to Compile and Loop through a PSO Problem

```
# SWARM OPTIMIZATION FUNCTION
# Parameters:
# ---- numDimensions: The number of dimension in the schwefel function
# ---- swarmSize: The number of particles within the swarm
# ---- intertiaWeight: The weight assigned to the intertia component of the velocity eq.
# ---- phi1: Cognitive weight of the velocity equation. Note phi1 + phi2 <= 4 # ---- phi2: Social weight of the velocity equation. Note phi1 + phi2 <= 4
# ---- totalIterations: The total number of iterations before stopping (Stopping criterion)
# ---- method: Can use 'local' or 'global' best methods.

# ---- filename: Name of CSV file to export to working directory.

# ---- If using 2D and 'global' best method, will export CSV.

# ---- Reason for exporting is to read in data to R for plotting iterations

# ---- printDims: Print the dimensions' value or not
# -----
def swarmOptimizationSchwefel( # ------ Defaults ------
    numDimensions = 2, # number of dimensions of problem
    swarmSize = 5, # number of particles in swar

phi1 = 2, # Cognitive weight

phi2 = 2, # Social weight

intertiaWeight = 0.1, # Constant Inertia weighting value

totalIterations = 100, # Stopping criteria = the total number of iterations
    method = 'local', # 'local' or 'global' best function name
    filename = 'output', # Name of the ouput csv file to write first 5 iterations,
printDims = False # Print the dimension's value or not
    ):
     # Initialize to global best function by default
    functionToGetBest = getGlobalBest
     # If not using the global best, then switch to the local best method
     if method != 'global':
         functionToGetBest = getLocalBest
     # ------
     # INITIALIZE POSITION AND VELOCITY, and INITIAL BESTS
     # the swarm will be represented as a list of positions, velocities, values,
     # pBestPosition, and pBestPosition values
     # Note: position[0] and velocity[0] provides the position and velocity of particle 0;
     # position[1] and velocity[1] provides the position and velocity of particle 1; and so on.
     # Step 1: Initialize swarm and get the particles' and global best (and current position)
    position, velocity, pCurrFitValue, pBestPosition, pBestFitValue, glBestFitValue, glBestPosition = i
     # Create empty lists for holding the swarm iterations
    positionIterations = [] # Each particle's velocity
```

```
velocityIterations = [] # Each particle's position
gBestPositionIterations = [] # The current Global Best Position
# -----
# Main Loop
                 _____
for iteration in range(totalIterations):
   # Step 0: Keep track of each iterations/dimension for velocity, position, and current global be
   velocityIterations.append(velocity)
   positionIterations.append(position)
   gBestPositionIterations.append(glBestPosition)
   # Step 2: Update the velocity and position
   velocity, position = updateVelocityAndPosition(intertiaWeight, velocity, position,
                                             phi1, phi2, pBestPosition, glBestPosition,
                                             swarmSize, numDimensions)
   # Step 3: Recalculate the particle and global bests
   pCurrFitValue, pBestPosition, pBestFitValue = calculateParticleBests(position,
                                                                 swarmSize, numDimensions,
                                                                 pCurrFitValue,
                                                                 pBestPosition,
                                                                 pBestFitValue)
   # Step 4: Get the Global or local best (depends on chosen method) fitness value and position
   glBestFitValue, glBestPosition = functionToGetBest(pBestFitValue[:], pBestPosition[:], swarmSiz
# Global Best
# Finally, if using the local best method, get the absolute best from the local bests
if method == 'local':
   gBestFitValue, gBestPosition = getGlobalBest(glBestFitValue, glBestPosition, swarmSize)
else: # if not local best, then change the gl best is the global best
   gBestFitValue, gBestPosition = glBestFitValue, glBestPosition
# Print and Export
# -----
# Print the global (or local best) and each dimensions' position
displayGlobalBest(gBestFitValue, gBestPosition, numDimensions, printDims)
# If 2D, then write to a csv for plotting in R
writeIteratonsToCSV(numDimensions, filename, method, velocityIterations,
                 positionIterations, gBestPositionIterations, swarmSize)
```

#### 1.8 Part 1 (b) & (c) - Global Best Method

#### 1.8.1 1(b) 2D Global Swarm

Create a swarm of size 5 and solve the 2D Schwefel problem

```
# 1(b) Create a swarm of size 5 and solve the 2D Schwefel problem. -----
swarmOptimizationSchwefel(
   numDimensions = 2,
                            # number of dimensions of problem
                          # number of particles in swar
# Cognitive weight
   swarmSize
                  = 5,
                  = 2,
   phi1
                            # Social weight
   phi2
                  = 2,
                           # Constant Inertia weighting value
   intertiaWeight = 0.1,
   totalIterations = 100000,  # Stopping criteria = the total number of iterations
              = 'global', # 'local' or 'global' best function name
   method
   filename
                  = 'output', # Name of the ouput csv file to write first 5 iterations
                              \# Print the dimension's value or not
   printDims
                 = True
```

```
##
## Global Best Value: 0.0005
## For each [dimension], Global Best Position:
##
## | Dimension | Global Best |
## |-----|
## | 0 | 420.9120 |
## | 1 | 420.9941 |
## 0.0005125608778371316
```

#### 1.8.2 1(b) i. - iii. Iteration Table & Plot

# i. – ii. Record in table the first 5 positions and velocities of each particle. Highlight Global Best

```
library(tidyverse)
library(ggplot2)
library(readr) # Read CSV
library(knitr) # table printing
# Pull in the data from the swarm optimization (written in
# main loop)
swarmData <- read_csv("output.csv")</pre>
# Table holding the 5 iterations for the 5 particles
# Position and Velocity 1 & 2, as well indicator of global
# best
swarmDataTable <- swarmData %>%
    mutate(isGlobalBest = if_else(position1 == globalBest1 &
        position2 == globalBest2, "New Global Best!", "")) %>%
select(particle, iteration, starts_with("pos"), starts_with("vel"),
    isGlobalBest)
# Print the table
kable(swarmDataTable)
```

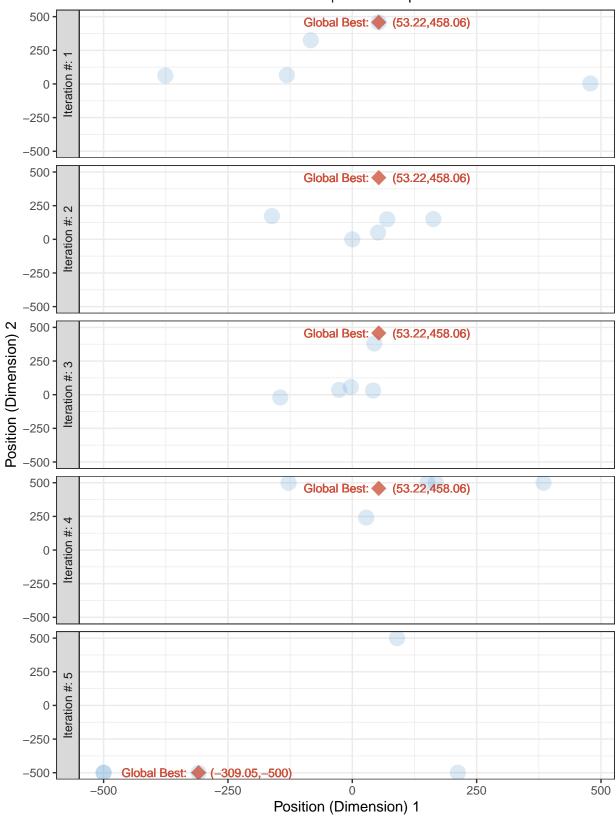
particle	iteration	position1	position2	velocity1	velocity2	is Global Best
0	1	-83.3801	325.2065	-0.9797	-0.4027	
1	1	-131.5883	66.0082	-0.6127	-0.6766	
2	1	-375.7331	62.0785	-0.1341	-0.6513	
3	1	53.2211	458.0648	-0.2902	-0.8174	New Global Best!
4	1	478.6400	3.9354	-0.1758	-0.7037	
0	2	51.8027	50.4383	-31.5775	375.6448	
1	2	70.1557	148.8914	-61.4326	214.8996	
2	2	162.9646	150.3870	-212.7685	212.4655	
3	2	-0.0290	-0.0817	53.1921	457.9830	
4	2	-161.6524	172.4728	316.9876	176.4082	
0	3	-3.0910	56.7402	48.7117	107.1785	
1	3	-144.7555	-20.5850	-74.5997	128.3064	
2	3	-26.4395	35.7205	136.5252	186.1075	
3	3	44.2005	380.3201	44.1715	380.2383	
4	3	41.8069	31.0758	-119.8455	203.5486	
0	4	151.1079	500.0000	148.0169	478.7985	
1	4	384.5587	500.0000	239.8032	479.4150	
2	4	167.7929	500.0000	141.3534	244.5300	
3	4	27.9970	241.2485	72.1975	-496.5454	
4	4	-128.0547	500.0000	-86.2478	440.2385	
0	5	-309.0507	-500.0000	-157.9428	0.0000	New Global Best!
1	5	-500.0000	-500.0000	-115.4413	0.0000	
2	5	-500.0000	-500.0000	-332.2071	0.0000	

particle	iteration	position1	position2	velocity1	velocity2	isGlobalBest
3	5	90.0355	500.0000	118.0325	-321.1262	
4	5	212.3308	-500.0000	84.2761	0.0000	

#### iii Plot the first 5 positions of each of the 5 particles

```
require(ggplot2)
# Plot the first 5 iterations with each particle's position
iterPlot <- swarmData %>%
  ggplot() +
  # Points for the Swarm
  geom_point(aes(x = swarmData$position1,
            y = swarmData$position2),
             alpha = 0.2, color = 'steelblue3', size = 5) +
  # # Red Point for the global best
  geom_point(aes(x = swarmData$globalBest1,
                y = swarmData$globalBest2),
             alpha = 0.25,
             color = 'tomato3', size = 5, shape = 18) +
  # # Label for the global best
  geom_text(label = paste0('
                                Global Best:
                           round(swarmData$globalBest1, 2), ',',
                           round(swarmData$globalBest2, 2),')'),
            aes(x = swarmData$globalBest1,
               y = swarmData$globalBest2),
           color = 'tomato3',
            # alpha = 0.9,
           size = 3) +
  # Facet by Iteration
  facet_grid(rows = vars(paste0('Iteration #: ', iteration)),
            switch = 'y') +
  labs(title = 'Particle Swarm Optimization',
      subtitle = paste0('Shows the first ', max(swarmData$iteration),
                         ' iterations of the PSO | Daniel Carpenter'),
      x = 'Position (Dimension) 1',
      y = 'Position (Dimension) 2') +
  theme_bw() # a theme
print(iterPlot)
```

# Particle Swarm Optimization Shows the first 5 iterations of the PSO | Daniel Carpenter



#### 1.8.3 1(c) 200D Global Swarm w/Various Param

Create a swarm to best and solve the 200D Schwefel problem

- Please note I did 50 iterations as stopping criteria simply for the sake a running to completion
- My computer cannot handle such large loads. It is likely that higher number of iterations would lead to closer to the known optima
- You can see that the 2D performs very well, as it does with increased dimensions (just not 200).

```
# 1(c) create a swarm to best and solve the 200D Schwefel problem. ------
# Try different swarm sizes, inertial weights, and values for phi1 and phi2

# 1(c, i) Base 200D
s1 = swarmOptimizationSchwefel(
    numDimensions = 200,  # number of dimensions of problem
    swarmSize = 5,  # number of particles in swar
    phi1 = 2,  # Cognitive weight
    phi2 = 2,  # Social weight
    intertiaWeight = 0.1,  # Constant Inertia weighting value
    totalIterations = 50,  # Stopping criteria = the total number of iterations
    method = 'global', # 'local' or 'global' best function name
    printDims = False  # Print the dimension's value or not
    )

# 1(c, ii.) 200D, changes: Swarm from 5 to 10
```

## Global Best Value: 76347.0722

## ## Global Best Value: 75781.8518

```
method
                    = 'global', # 'local' or 'global' best function name
    printDims
                                # Print the dimension's value or not
                     = False
# 1(c, iv.) 200D, changes: interia weight = 0.9
##
## Global Best Value:
                           75998.1002
s4 = swarmOptimizationSchwefel(
    numDimensions = 200, # number of dimensions of problem
                 = 10, # number of particles in swar
= 1, # Cognitive weight
= 3, # Social weight
    swarmSize
    phi1
    phi2
   intertiaWeight = 0.9,  # Constant Inertia weighting value
totalIterations = 50,  # Stopping criteria = the total number of iterations
    method = 'global', # 'local' or 'global' best function name
    printDims
                  = False  # Print the dimension's value or not
    )
# 1(c, v.) 200D, changes: phi1: 3, phi2: 1
##
## Global Best Value:
                           77573.9291
s5 = swarmOptimizationSchwefel(
    numDimensions = 200,
                               # number of dimensions of problem
    swarmSize = 10,
                              # number of particles in swar
# Cognitive weight
    phi1
                    = 3,
              = 1,
                               # Social weight
   phi2
   intertiaWeight = 0.9, # Constant Inertia weighting value
totalIterations = 50, # Stopping criteria = the total number of iterations
    method = 'global', # 'local' or 'global' best function name
                   = False # Print the dimension's value or not
    printDims
    )
# Compile the results
## Global Best Value:
                          74347.6093
swarmResults = [s1, s2, s3, s4, s5]
print('\n----- Question 1(c) Results of Swarm Runs -----')
##
## ----- Question 1(c) Results of Swarm Runs -----
for result in range(0, len(swarmResults)):
    print('Result', result, ':\t', swarmResults[result])
```

```
## Result 0 : 76347.07221466661

## Result 1 : 75781.85183383343

## Result 2 : 75998.1002495657

## Result 3 : 77573.92907934781

## Result 4 : 74347.60932545959
```

#### 1.9 Part 1 (d) & (e) - Local Best Method

#### 1.9.1 Reference: Local Swarm Calculation and Logic

Uses a Ring Topology for Local Best Method

• Note this function already defined but placed here for ease of referencing

```
# LOCAL MIN VALUE AND POSITION SEARCH FUNCTION
# Topology: Ring structure with n neighbors (default 2)
# Returns the 2 element list of lists with the each particle's local best within neighborhood
# ---- [0] min value and
# ---- [1] associate position of
# Can change numParticlesInNbrhood to consider more or less in particle's neighborhood
# ------
def getLocalBest(fitnessValues, positions, swarmSize,
                numParticlesInNbrhood = 2): # Number of particles to compare to for local best
   lBestFitValue = [] # will hold the best VALUE of the n surrounding particles, for each particle
   lBestPosition = [] # will hold the best POSITION of the n surrounding particles, for each particle
    # For every particle in the swarm, (starting at n less than index 0)
   for particle in range(-numParticlesInNbrhood, swarmSize - numParticlesInNbrhood):
       # Identify the two neighbors fitness value of this particle,
       # which are the two precedng particles
       personalBestNeighbor1 = fitnessValues[particle]
       personalBestNeighbor2 = fitnessValues[particle + 1]
       # Identify the lowest fitness value of this particle's the two preciding neighbors
       minNeighValue = min(personalBestNeighbor1, personalBestNeighbor2)
       # Store the index of the particle
       minNeighIndex = fitnessValues.index(minNeighValue)
       # Store the particle's best neighbors fitness value and position
       lBestFitValue.append(fitnessValues[minNeighIndex])
       lBestPosition.append(positions[minNeighIndex])
    # Returns a list of particles and the min of their n best fit. valued neighbors
   return[lBestFitValue, lBestPosition]
```

#### 1.9.2 1(d) 2D Local Swarm

Implement a PSO algorithm that uses the local best with ring topology in place of the global best. See above logic and explanation

```
# 1(d) Create a swarm of size 5 and solve the 2D Schwefel problem. ------
# Implement a PSO algorithm that uses the "local best" in place of the
# qlobal best and explain which topology you are using.
# Topology: Ring structure with 2 neighbors
swarmOptimizationSchwefel(
   numDimensions = 2,
                            # number of dimensions of problem
   swarmSize = 5,
                            # number of particles in swar
                 = 2,
                            # Cognitive weight
   phi1
                = 2,
                            # Social weight
   phi2
   intertiaWeight = 0.1, # Constant Inertia weighting value
   totalIterations = 100000,  # Stopping criteria = the total number of iterations
   method = 'local', # 'local' or 'global' best function name
                 = 'output', # Name of the ouput csv file to write first 5 iterations
   filename
   printDims
                  = True
                            # Print the dimension's value or not
   )
```

```
##
## Global Best Value: 0.0004
## For each [dimension], Global Best Position:
##
## | Dimension | Global Best |
## |-----|
## | 0 | 420.9179 |
## | 1 | 420.9590 |
## 0.0003635313285030861
```

#### 1.9.3 1(e) 200D Local Swarm w/Various Params.

Solve the 200D Schwefel problem as best as possible

```
# 1(e) create a swarm to best and solve the 200D Schwefel problem. ------
# Same as 1(c) parameters, just using 'local' now
# Topology: Ring structure with 2 neighbors
# 1(e, i) Base 200D
s1e = swarmOptimizationSchwefel(
       numDimensions = 200,  # number of dimensions of problem
      swarmSize = 5,  # number of particles in swar

phi1 = 2,  # Cognitive weight

phi2 = 2,  # Social weight

intertiaWeight = 0.1,  # Constant Inertia weighting value

totalIterations = 50,  # Stopping criteria = the total number of iterations
       method = 'local', # 'local' or 'global' best function name
       printDims = False # Print the dimension's value or not
       )
# 1(e, ii.) 200D, changes: Swarm from 5 to 10
##
## Global Best Value:
                               76065.0451
s2e = swarmOptimizationSchwefel(
                                   # number of dimensions of problem
    numDimensions = 200,
                                   # number of particles in swar
# Cognitive weight
                       = 10,
    swarmSize
                       = 2,
    phi1
    phi2 = 2,  # Social weight
intertiaWeight = 0.1,  # Constant Inertia weighting value
totalIterations = 50,  # Stopping criteria = the total number of iterations
method = 'local',  # 'local' or 'global' best function name
    printDims = False
                                     # Print the dimension's value or not
    )
# 1(e, iii.) 200D, changes: phi1: 1, phi2: 3
## Global Best Value:
                               77721.3454
s3e = swarmOptimizationSchwefel(
    numDimensions = 200, # number of dimensions of problem
                                # number of particles in swar
# Cognitive weight
# Social weight
                      = 10,
    swarmSize
                        = 1,
    phi1
    phi2
                        = 3,
    intertiaWeight = 0.1,  # Constant Inertia weighting value
totalIterations = 50,  # Stopping criteria = the total number of iterations
                  = 'local', # 'local' or 'global' best function name
    method
                                      # Print the dimension's value or not
    printDims
                      = False
# 1(e, iv.) 200D, changes: interia weight = 0.9
```

```
##
## Global Best Value: 75777.2461
s4e = swarmOptimizationSchwefel(
    numDimensions = 200, # number of dimensions of problem
    swarmSize = 10, # number of particles in swar

phi1 = 1, # Cognitive weight

phi2 = 3, # Social weight

intertiaWeight = 0.9, # Constant Inertia weighting value

totalIterations = 50, # Stopping criteria = the total number of iterations
    method = 'local', # 'local' or 'global' best function name
                    = False
    printDims
                                  # Print the dimension's value or not
    )
# 1(e, v.) 200D, changes: phi1: 3, phi2: 1
##
## Global Best Value:
                             75704.8377
s5e = swarmOptimizationSchwefel(
    numDimensions = 200,
                                   # number of dimensions of problem
                                 # number of particles in swar
# Cognitive weight
                      = 10,
    swarmSize
                      = 3,
    phi1
    phi2 = 1,  # Social weight
intertiaWeight = 0.9,  # Constant Inertia weighting value
totalIterations = 50,  # Stopping criteria = the total number of iterations
    method = 'local', # 'local' or 'global' best function name
    printDims
                    = False
                                   # Print the dimension's value or not
## Global Best Value:
                             75700.0155
swarmResultsE = [s1e, s2e, s3e, s4e, s5e]
print('\n----- Question 1(e) Results of Swarm Runs -----')
##
## ----- Question 1(e) Results of Swarm Runs -----
for result in range(0, len(swarmResultsE)):
    print('Result', result, ':\t', swarmResultsE[result])
## Result 0 : 76065.04505657653
## Result 1 : 77721.34535963654
## Result 2 :
                   75777.24606450333
## Result 3 : 75704.83774699294
## Result 4 : 75700.0155348131
```