

1. Experimental Results

1.1. Number of Individuals

Number of generations = 1000



Figure 1 : Number of Individuals is 5



Figure 2 : Number of Individuals is 10



Figure 3 : Default values

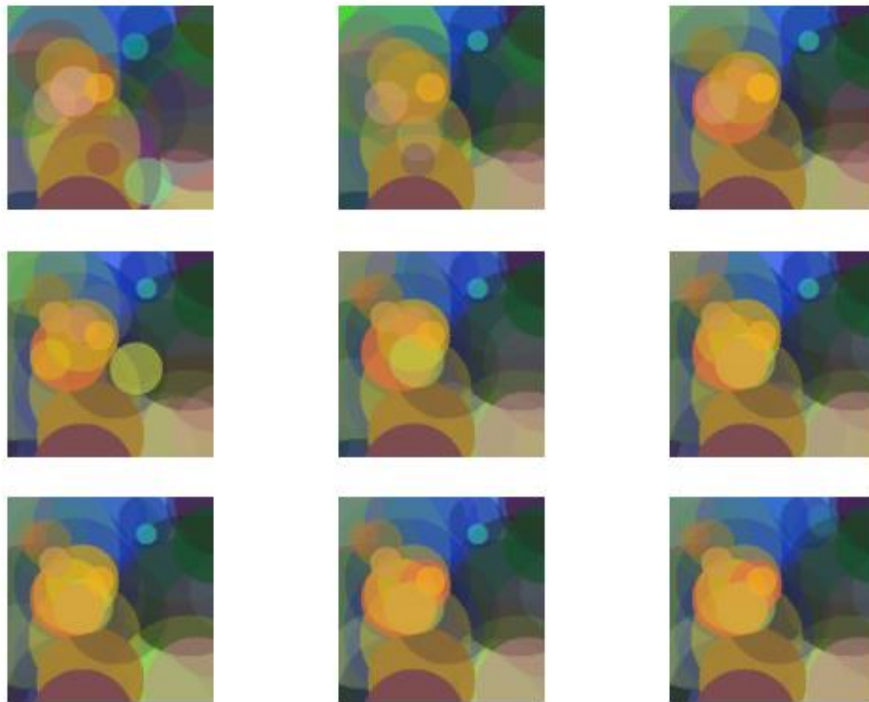


Figure 4 : Number of Individuals is 40

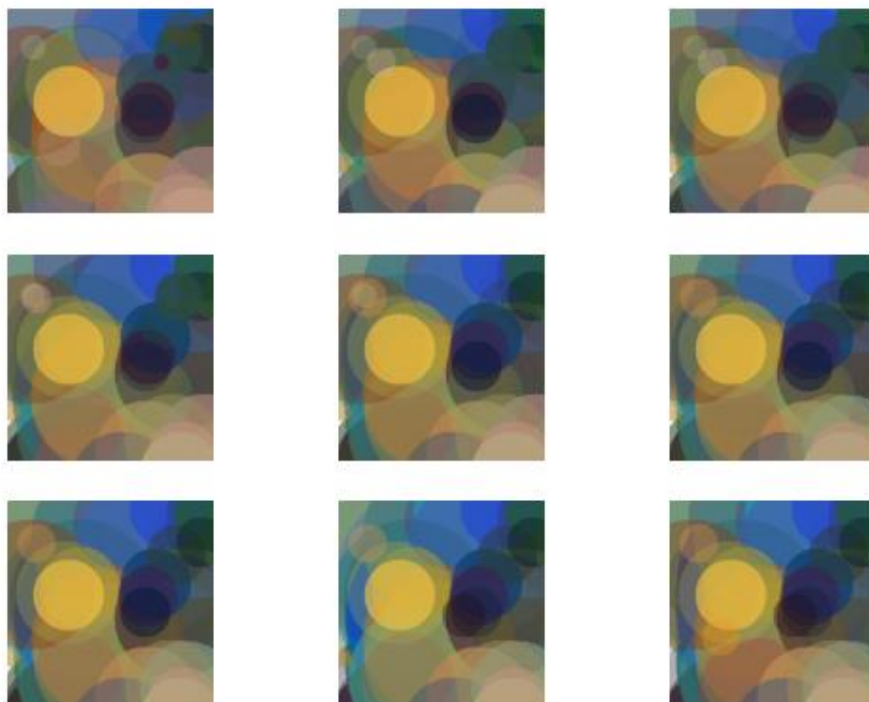


Figure 5 : Number of Individuals is 60

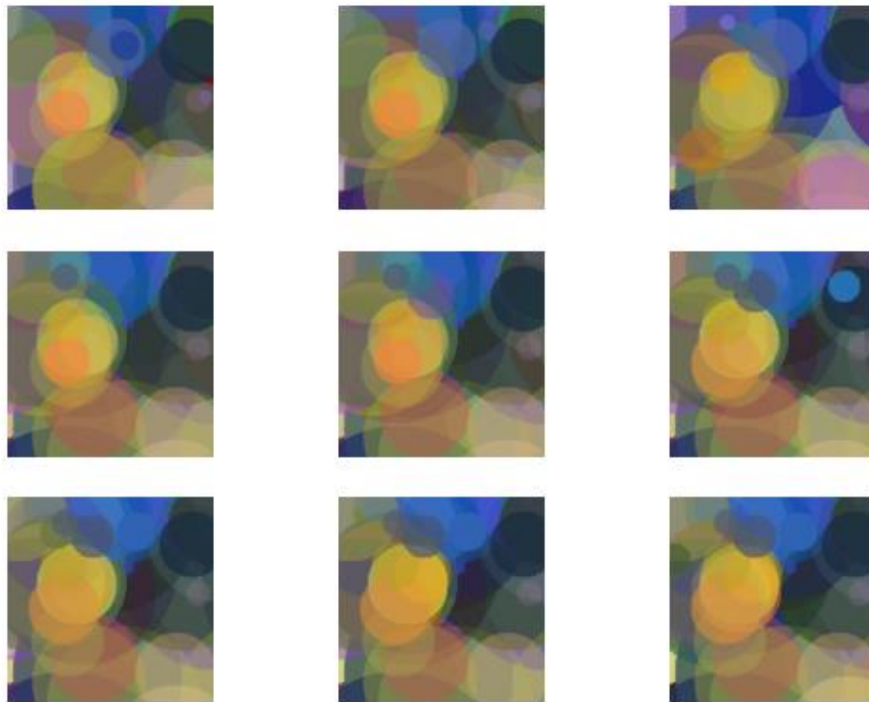
num_inds/5/iterations: 2000 - 10000



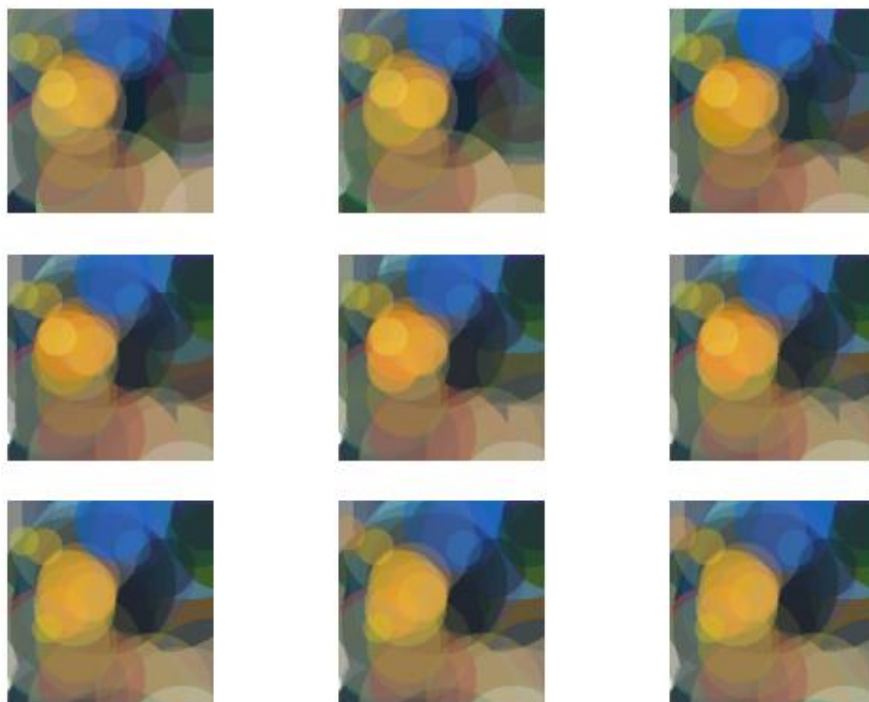
num_inds/10/iterations: 2000 - 10000



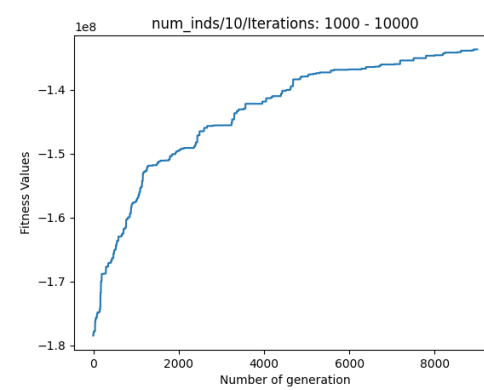
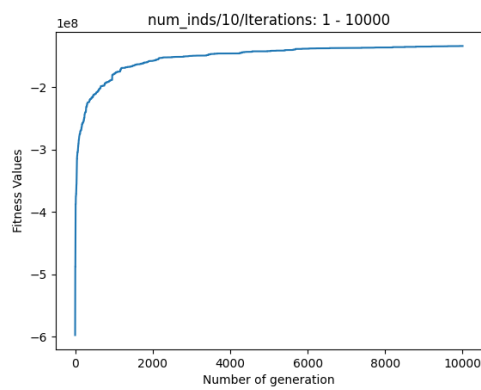
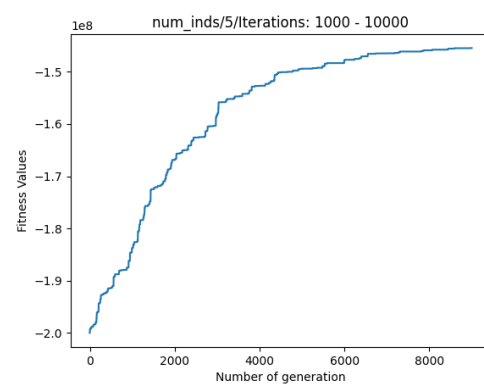
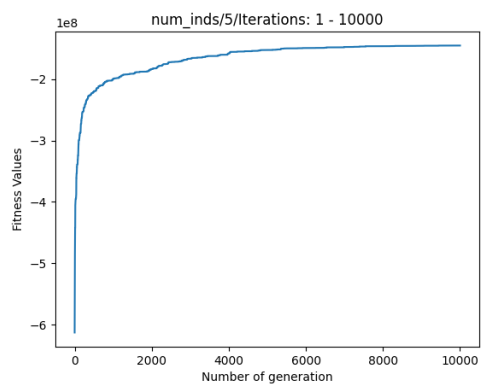
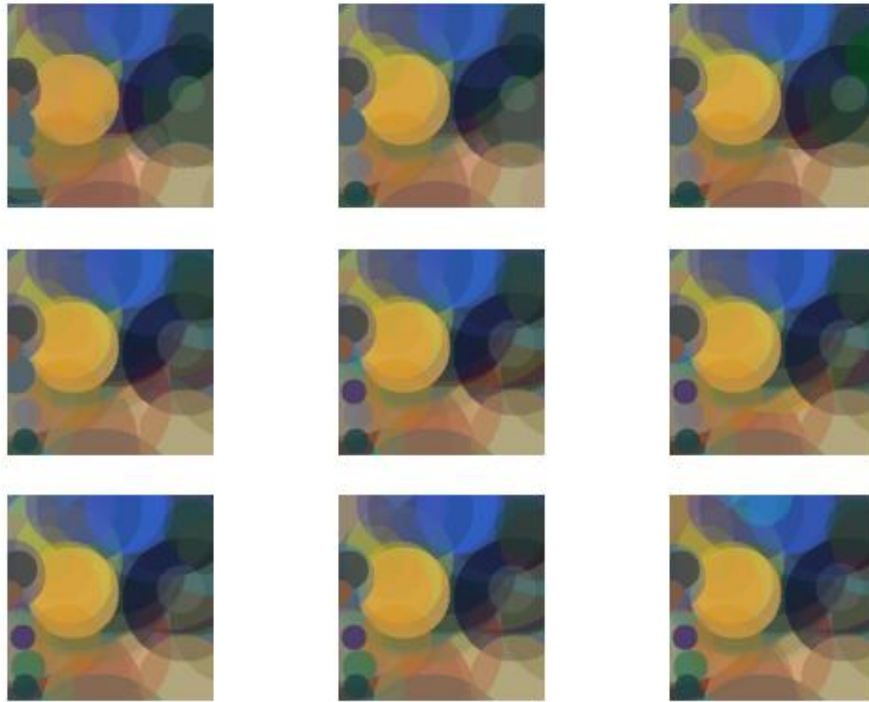
num_inds/20/Iterations: 2000 - 10000

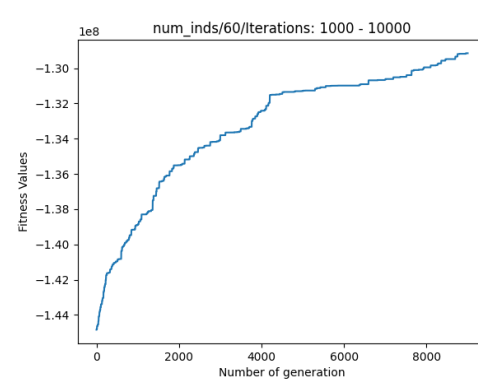
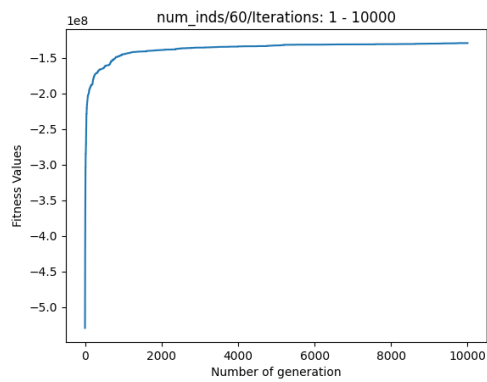
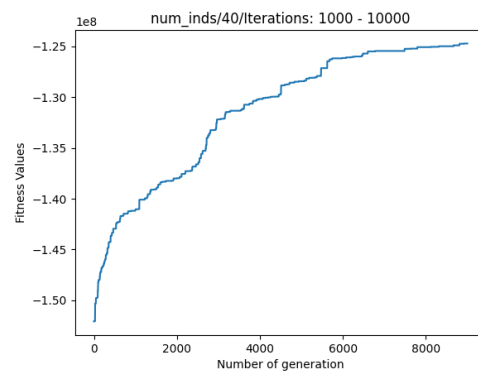
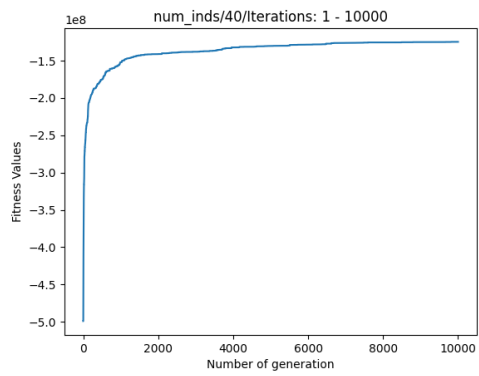
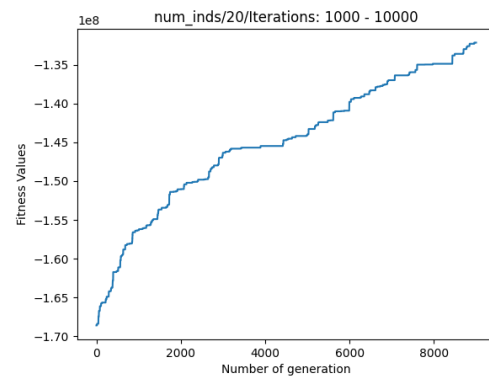
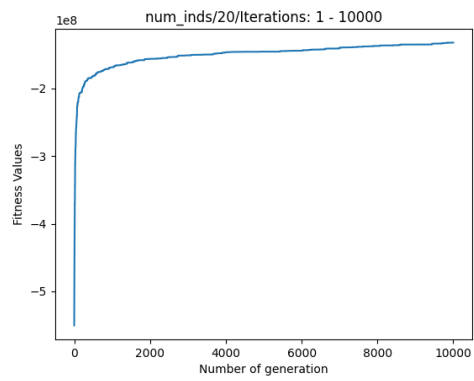


num_inds/40/Iterations: 2000 - 10000



num_inds/60/Iterations: 2000 - 10000





1.2. Number of Genes

Number of generations = 1000



Figure 6 : Number of Genes is 15



Figure 7 : Number of Genes is 30

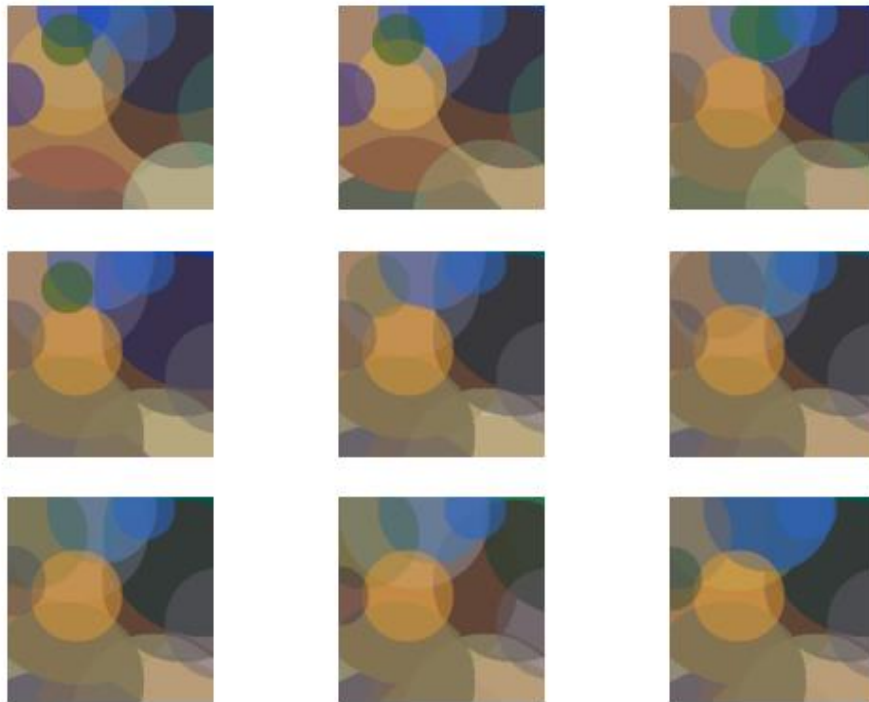


Figure 8: Number of Genes is 80

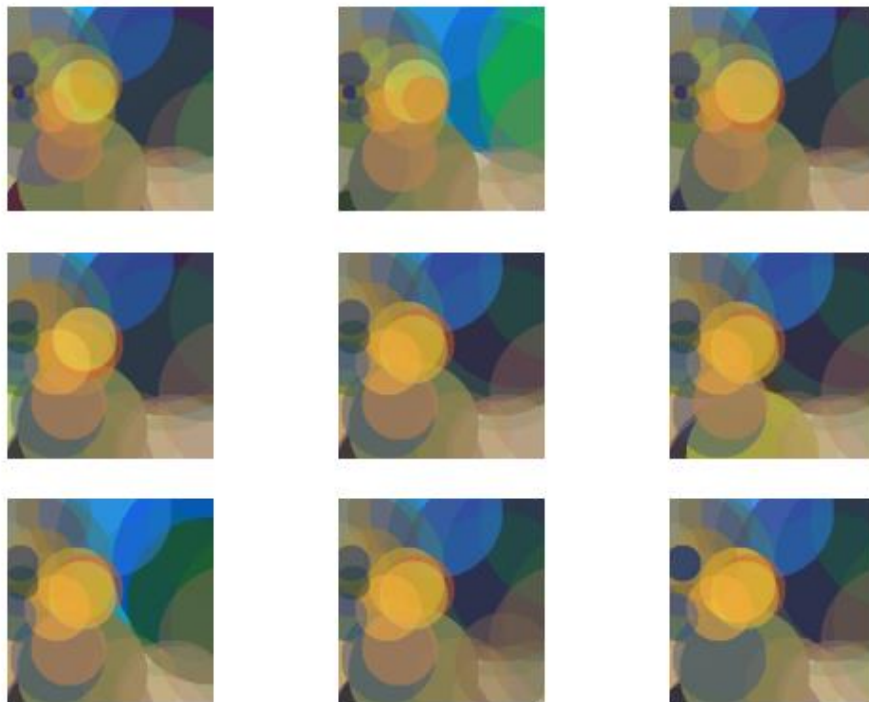


Figure 9 : Number of Genes is 120

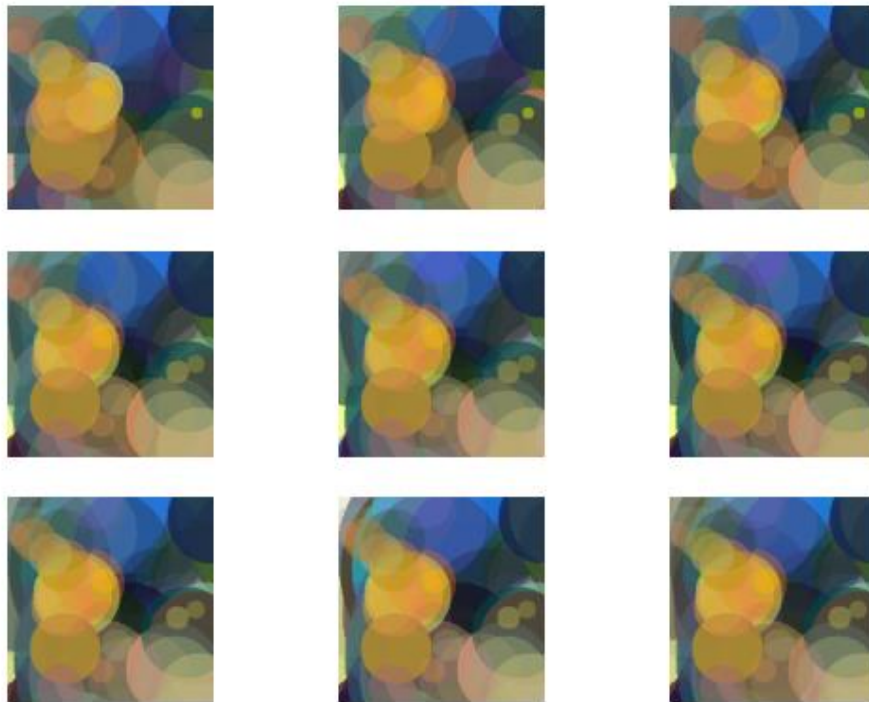
num_genes/15/Iterations: 2000 - 10000



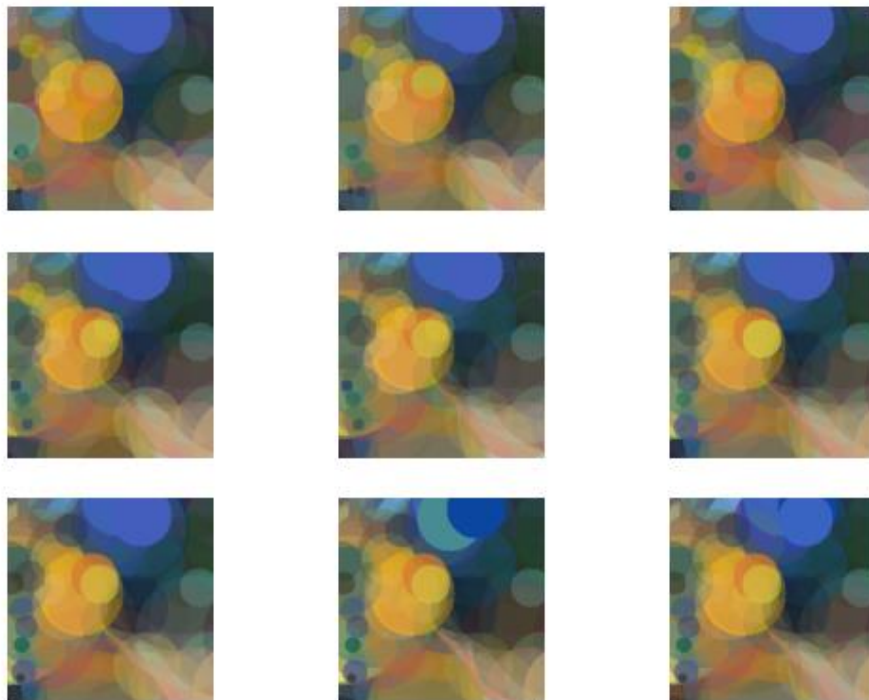
num_genes/30/Iterations: 2000 - 10000

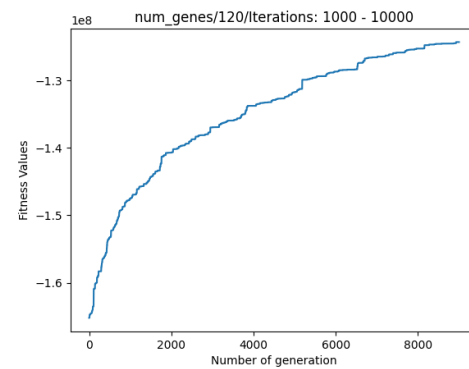
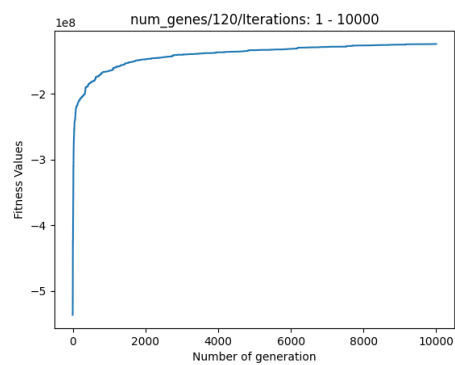
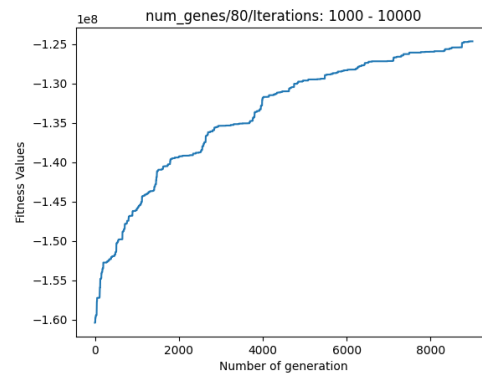
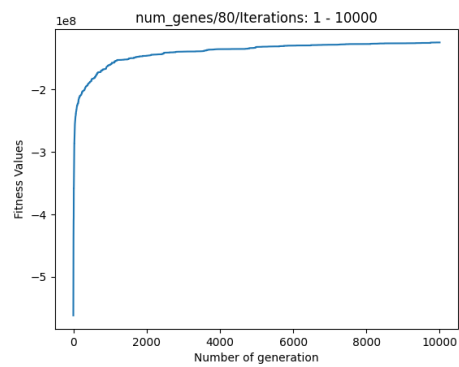
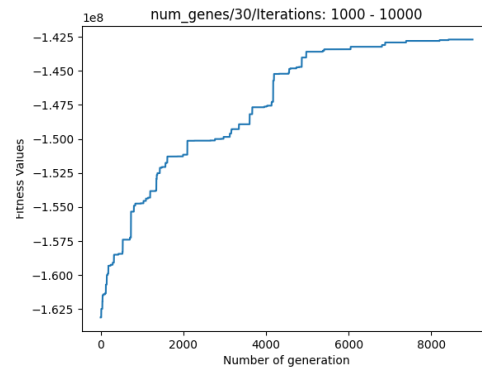
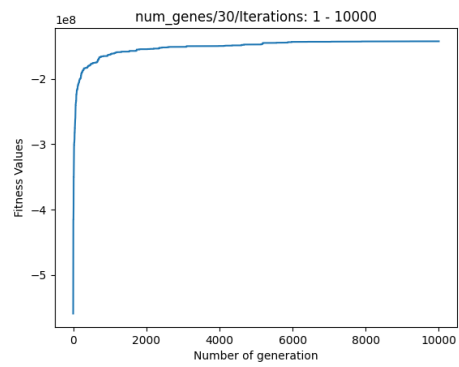
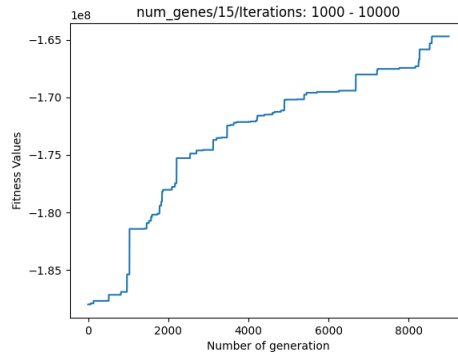
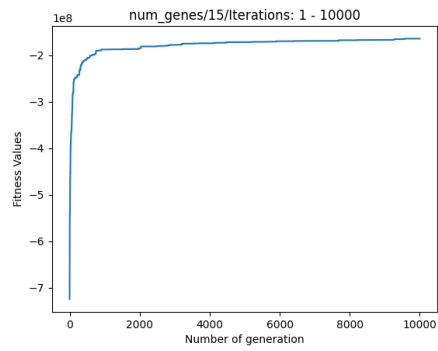


num_genes/80/Iterations: 2000 - 10000



num_genes/120/Iterations: 2000 - 10000





1.3. Tournament Size

Number of generations = 1000



Figure 10: Tournament Size is 2

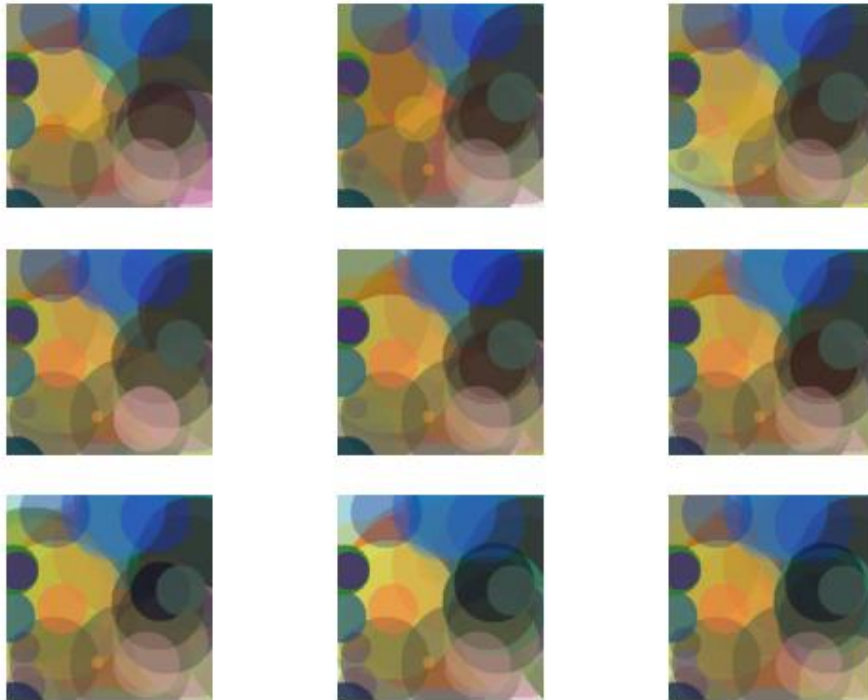


Figure 11 : Tournament Size is 8

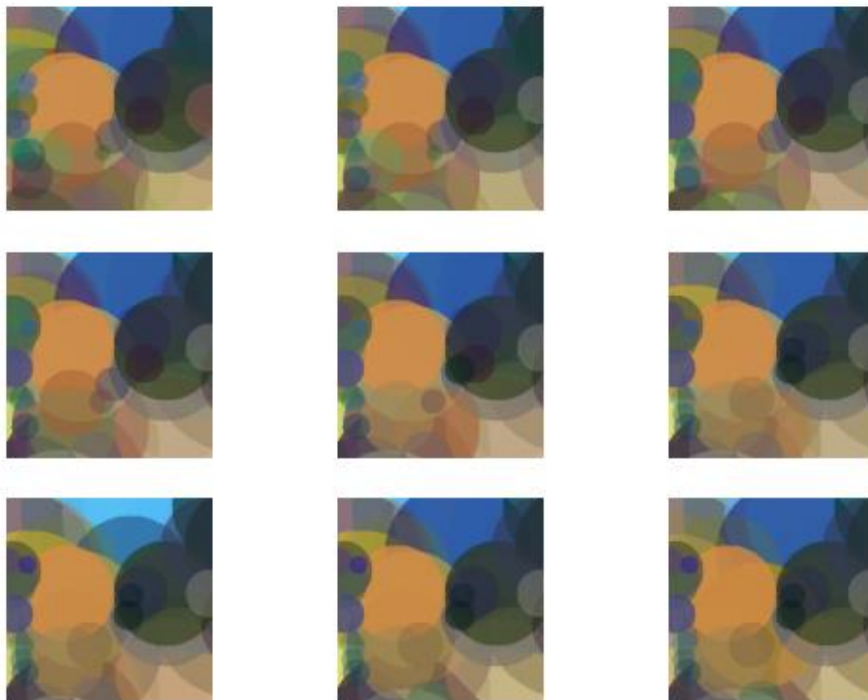


Figure 12 : Number of Genes is 16

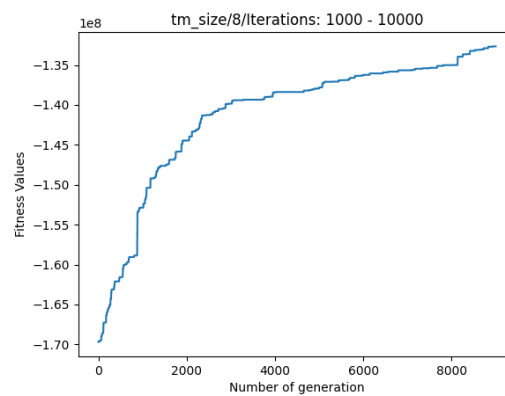
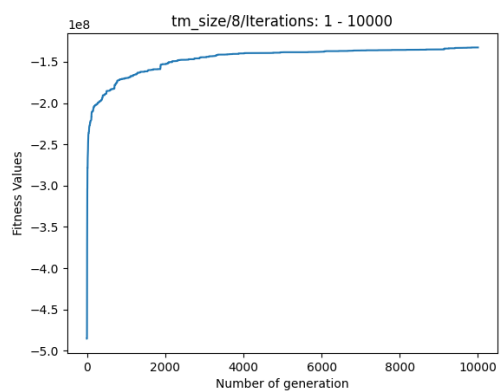
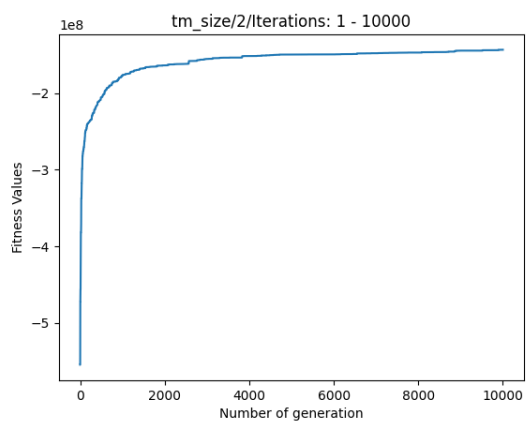
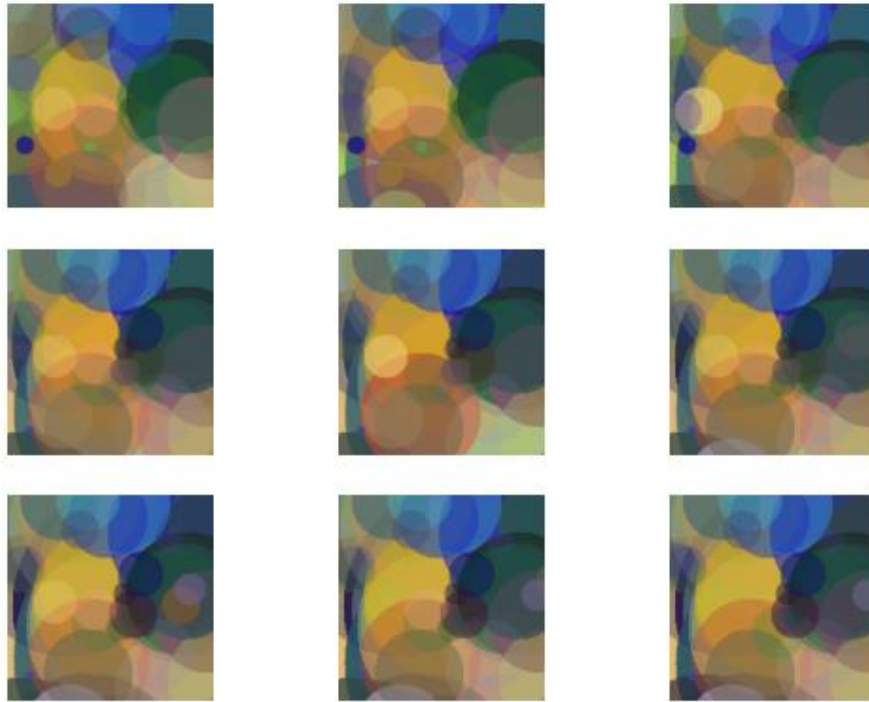
tm_size/2/Iterations: 2000 - 10000

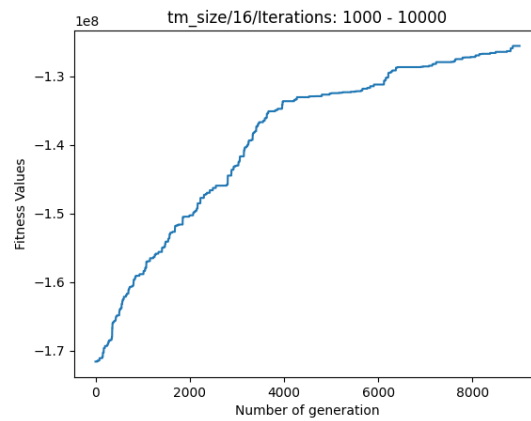
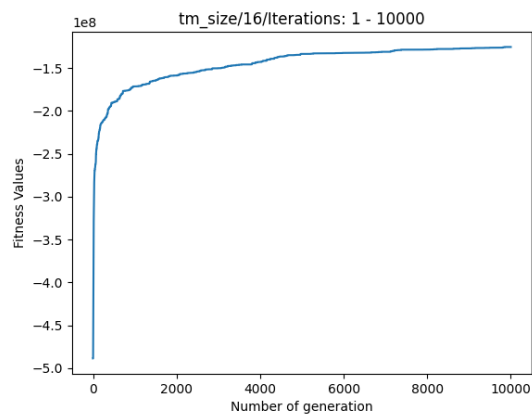


tm_size/8/Iterations: 2000 - 10000



tm_size/16/Iterations: 2000 - 10000





1.4. Number of Elites

Number of generations = 1000



Figure 13 : Number of Elits is 0.04

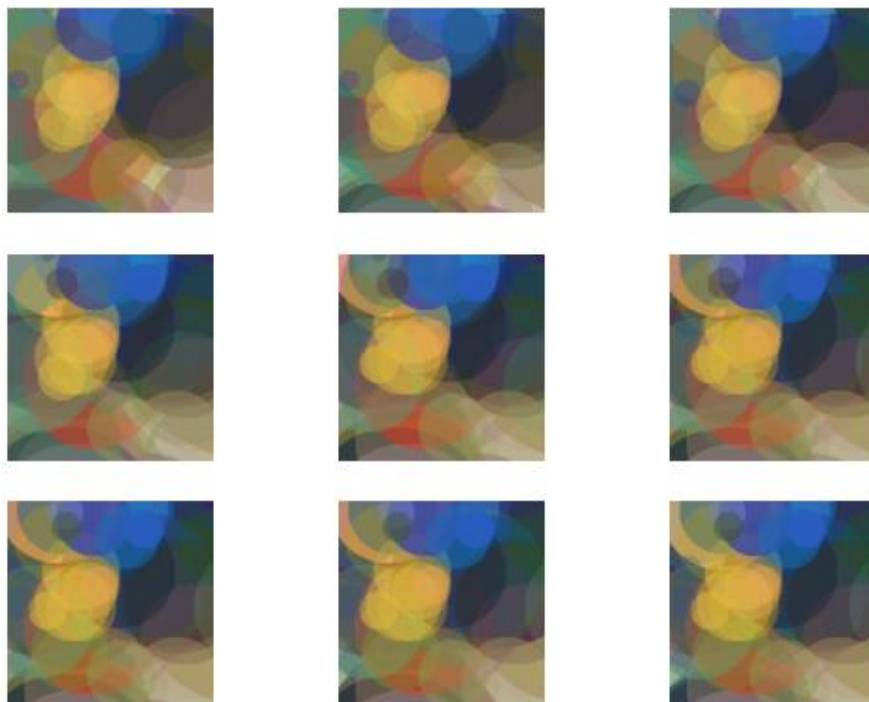


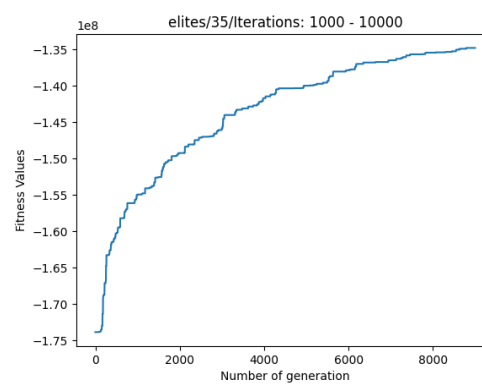
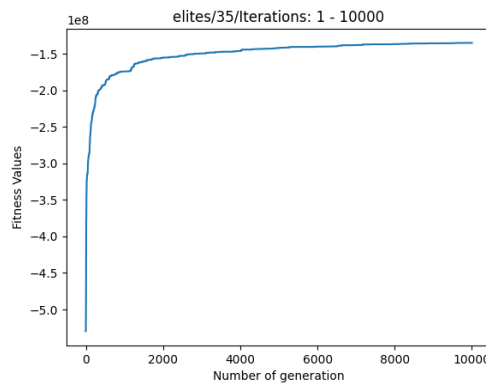
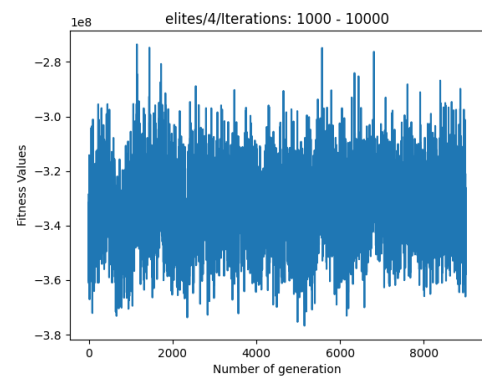
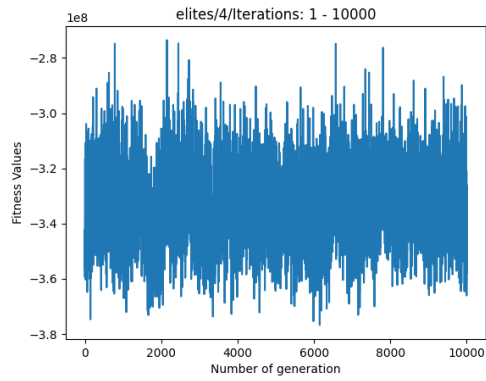
Figure 14 : Number of Elits is 0.35

elites/4/Iterations: 2000 - 10000



elites/35/Iterations: 2000 - 10000





1.5. Number of Parents

Number of generations = 1000



Figure 15 : Number of Parents is 0.15



Figure 16 : Number of Parents is 0.30

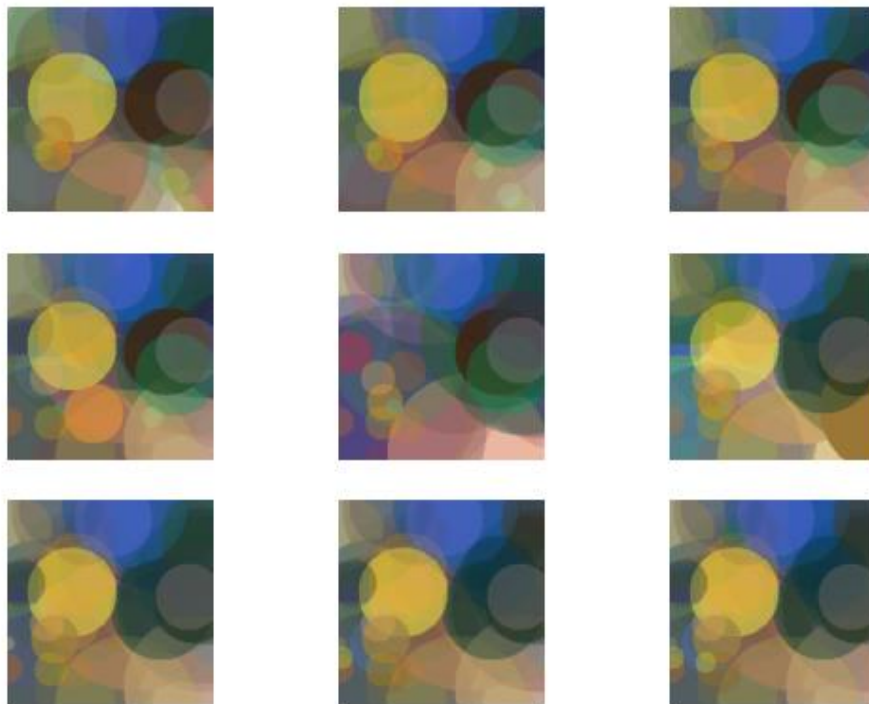


Figure 17: Number of Parents is 0.75

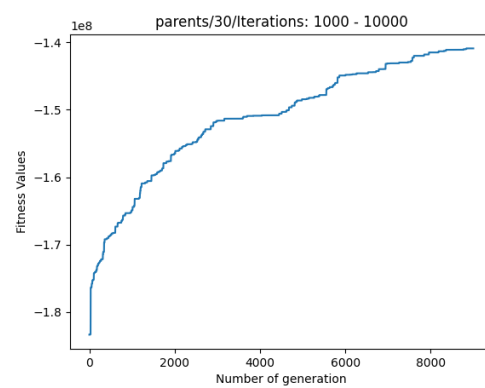
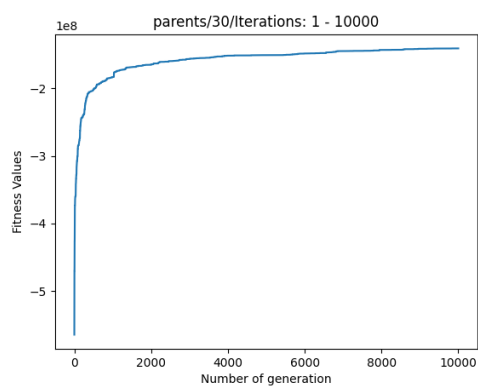
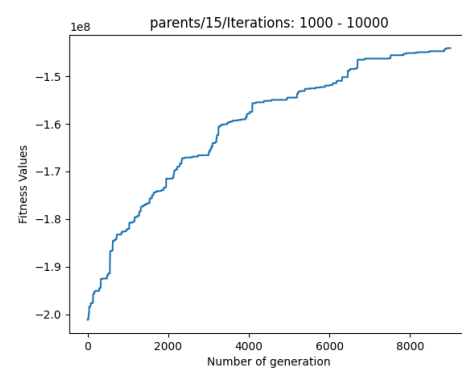
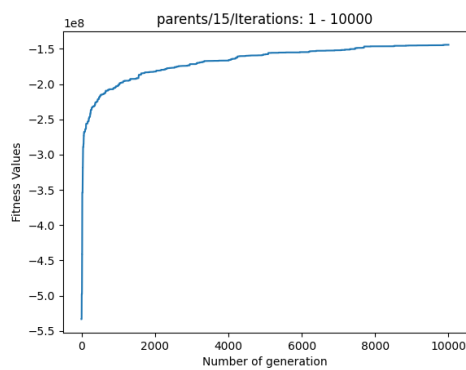
parents/15/Iterations: 2000 - 10000

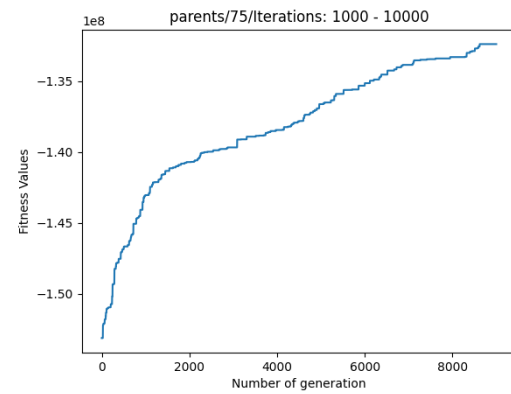
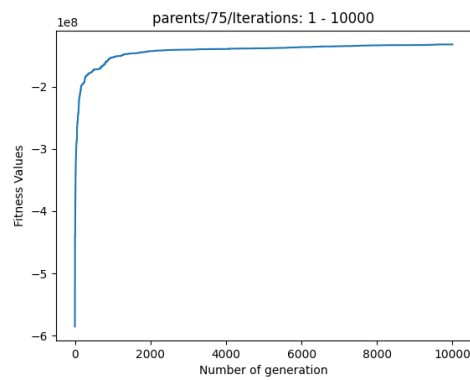


parents/30/Iterations: 2000 - 10000



parents/75/Iterations: 2000 - 10000





1.6. Mutation Probability

Number of generations = 1000



Figure 18 : Mutation Probability is 0.1

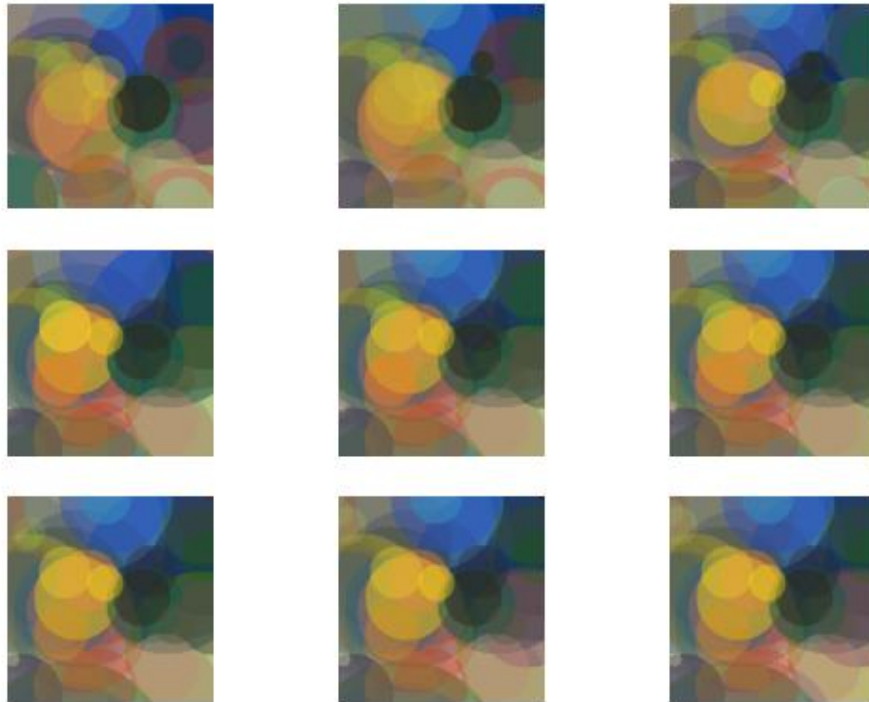


Figure 19 : Mutation Probability is 0.4

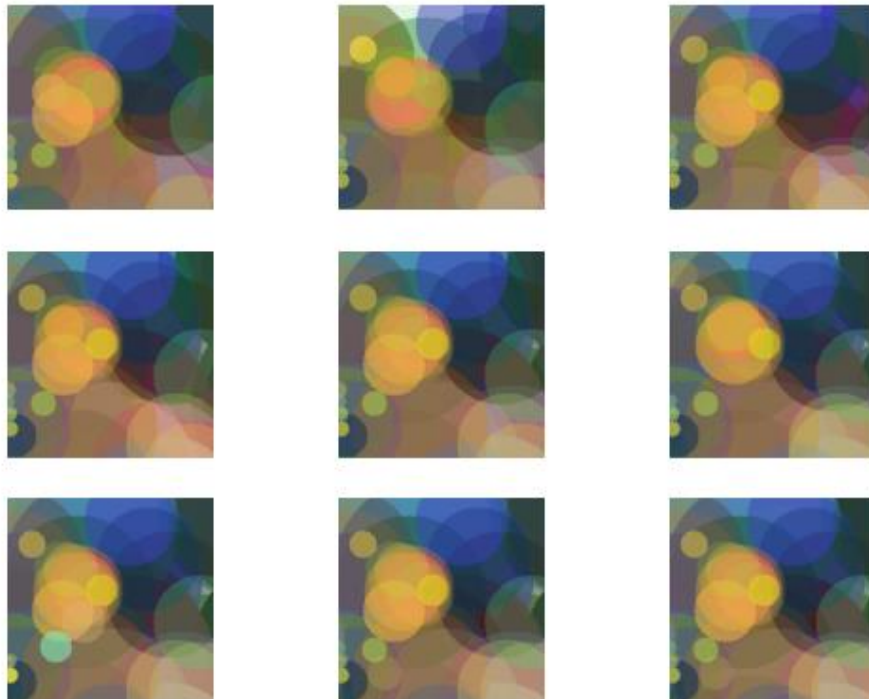


Figure 20 : Mutation Probability is 0.75

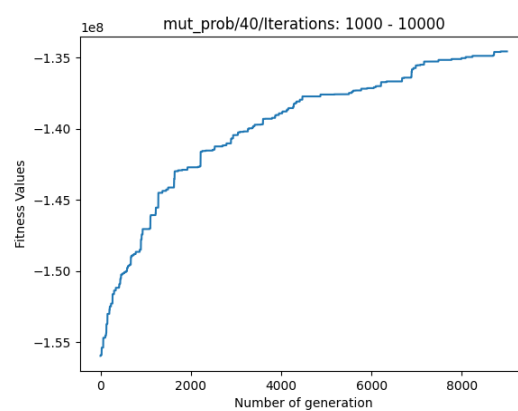
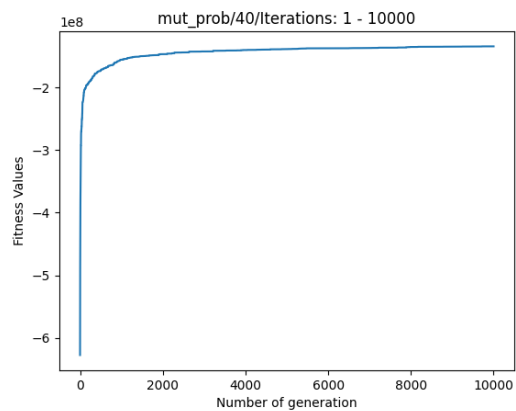
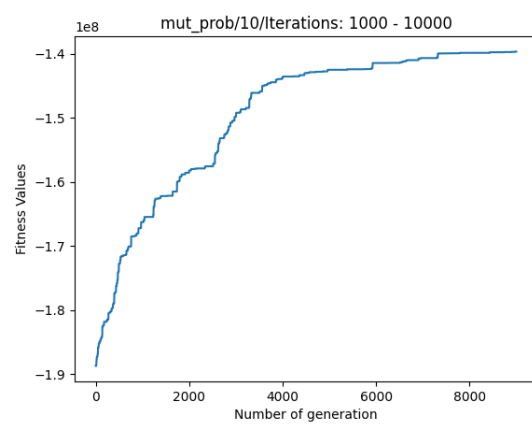
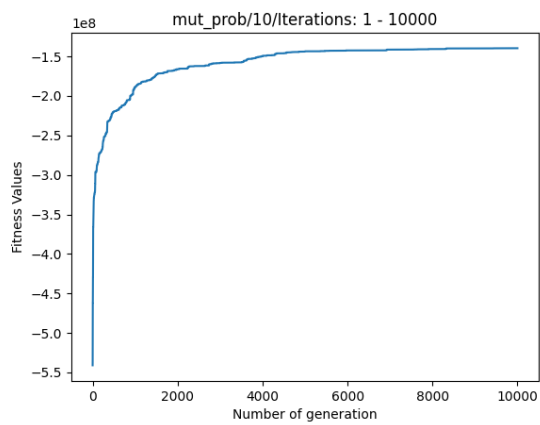
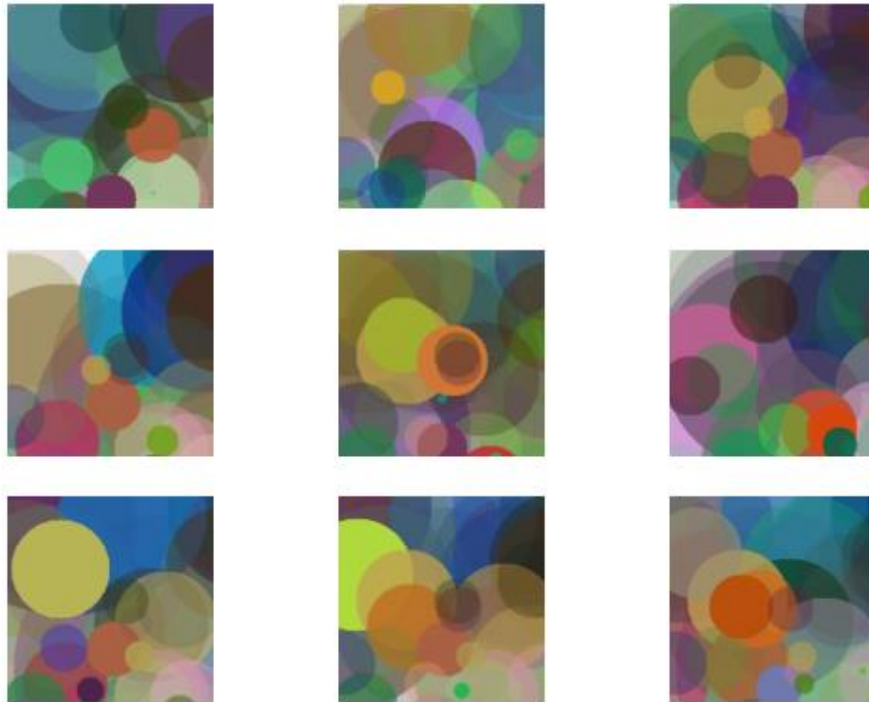
mut_prob/10/Iterations: 2000 - 10000

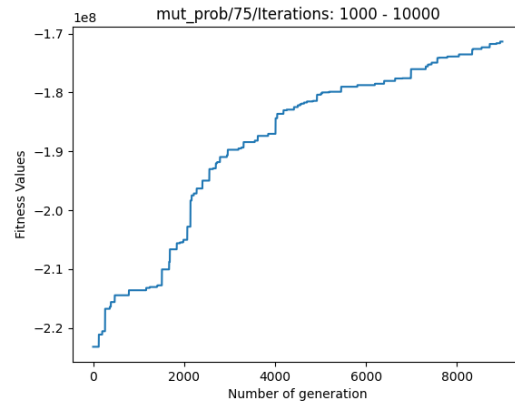
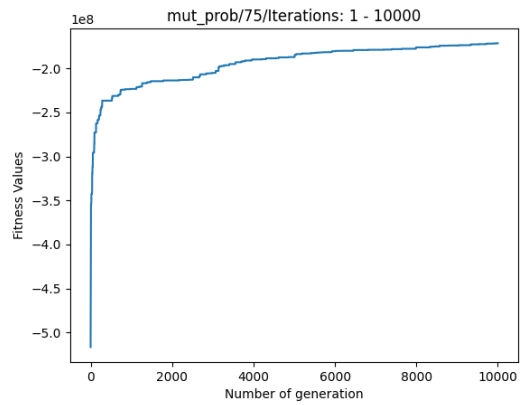


mut_prob/40/Iterations: 2000 - 10000



mut_prob/75/Iterations: 2000 - 10000



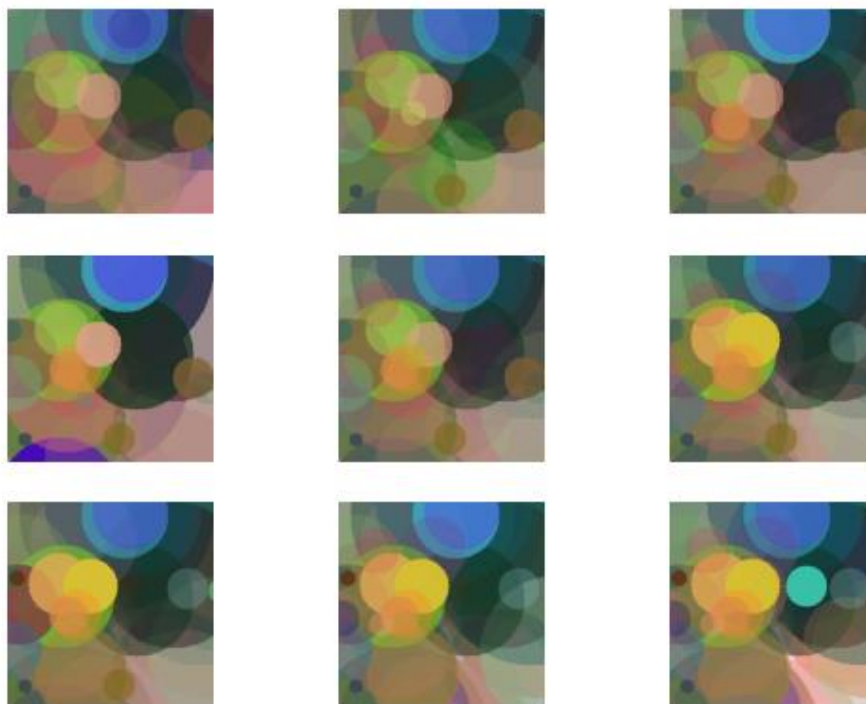


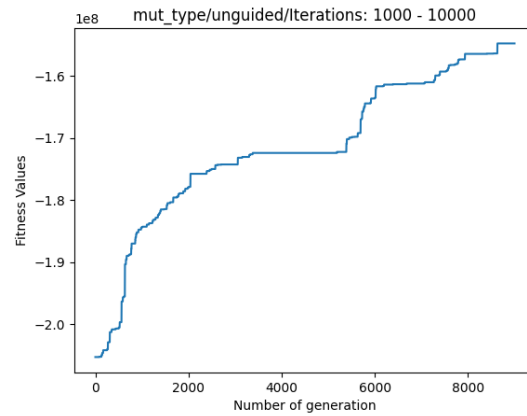
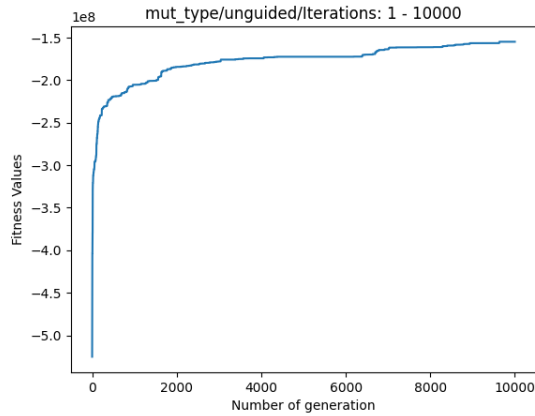
1.7. Mutation type



Figure 21 : Mutation Type is unguided

mut_type/unguided/Iterations: 2000 - 10000





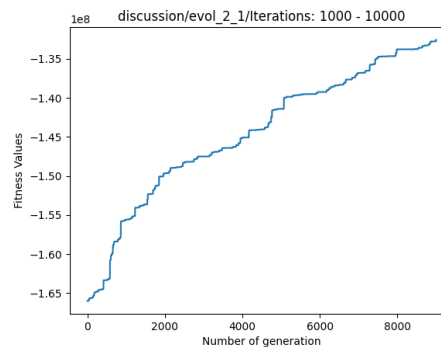
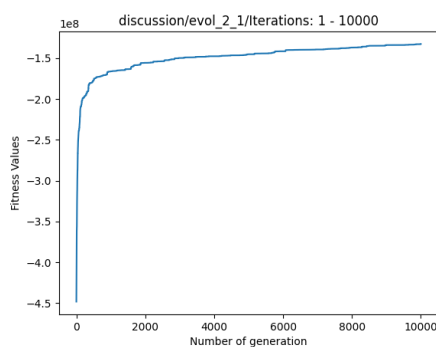
2. Discussions

2.1. Suggestion 1

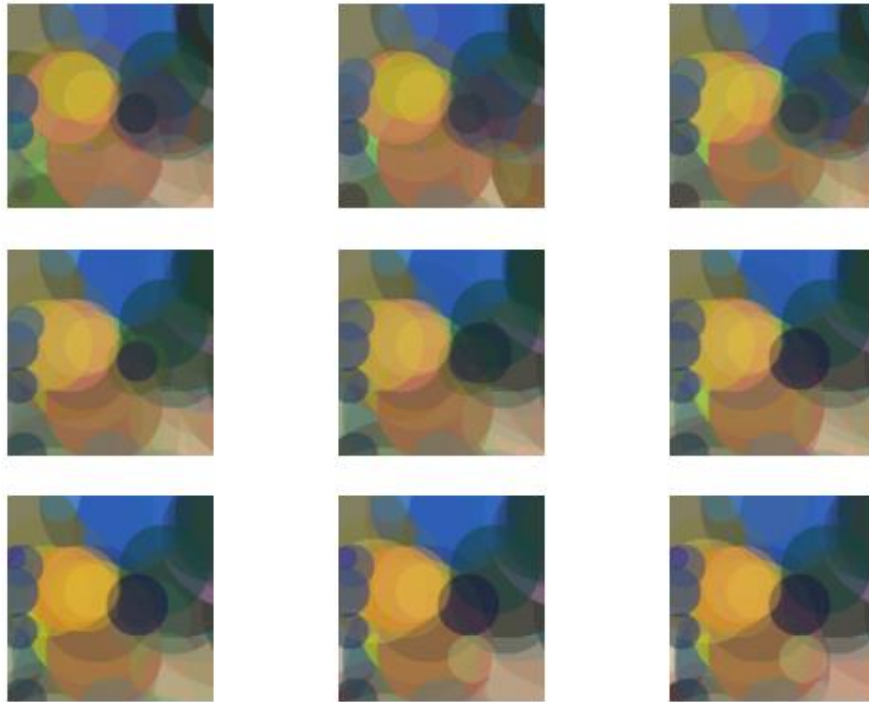
In this suggestion, I propose selecting the two best members from each family: a father, a mother, and two children. We ensure continuous progress and improvement with each iteration by choosing the most promising individuals from each family during the crossover.



Figure 22: 1000th generation of Suggestion 1



discussion/evol_2_1/iterations: 2000 - 10000

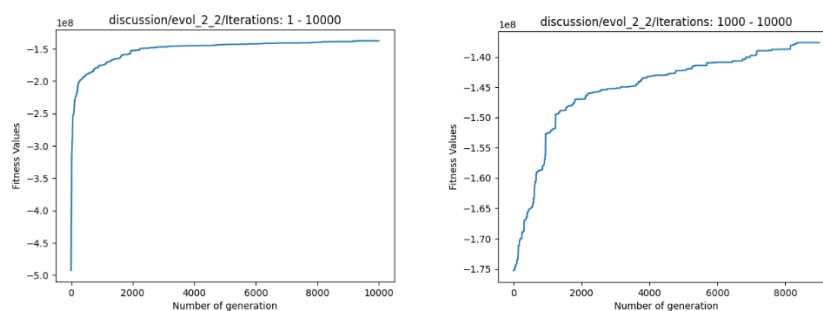


2.2. Suggestion 2

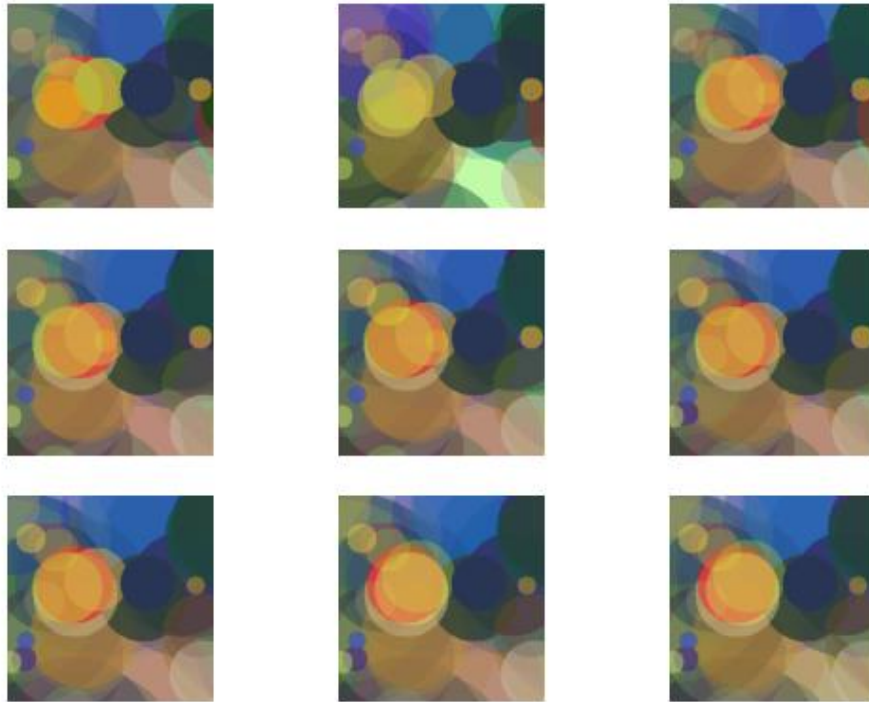
In this suggestion, we propose removing the last two individuals from the population during selection. These individuals have the lowest chances of improving themselves to become elites. By replacing them with new individuals, we provide fresh opportunities for progress and advancement within the population.



Figure 23: 1000th generation of Suggestion 2



discussion/evol_2_2/iterations: 2000 - 10000



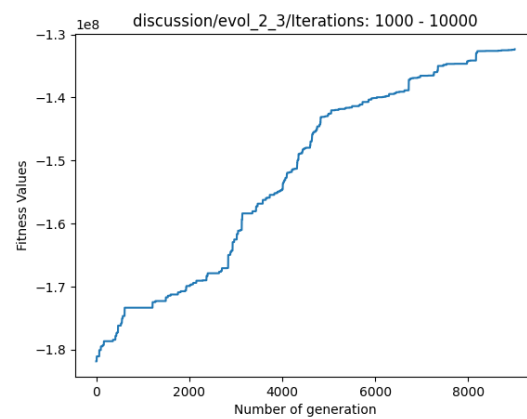
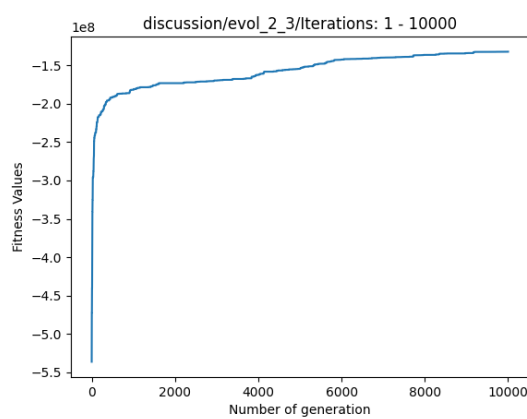
2.3. Suggestion 3

We propose adjusting mutation types and probabilities based on generations in this suggestion. Initially, for the first 5000 generations, we increase the mutation probability to prioritize the rapid improvement of individuals. However, after 5000 generations, we shift to a guided mutation type to ensure the preservation of successful individuals while still introducing variability.

```
def adjust_mut_parameters(self, iteration):  
    if iteration < 1000:  
        self.mutation_type = 'unguided'  
        self.mutation_prob = 0.8  
    elif iteration > 5000:  
        self.mutation_type = 'guided'  
        self.mutation_prob = 0.5
```



Figure 24 : 1000th generation of Suggestion 3



discussion/evol_2_3/Iterations: 2000 - 10000



HW2 code

```
import cv2
import numpy as np
import random as rnd
import json
import matplotlib.pyplot as plt
from copy import deepcopy
source_image = cv2.imread("painting.png")

WIDTH, HEIGHT = source_image.shape[0], source_image.shape[1]

num_inds = [5, 10, 20, 40, 60]
num_genes = [15, 30, 50, 80, 120]
tm_size = [2, 5, 8, 16]
frac_elites = [0.04, 0.2, 0.35]
frac_parents = [0.15, 0.3, 0.6, 0.75]
mutation_prob = [0.1, 0.2, 0.4, 0.75]
mutation_type = ["guided", "unguided"]
num_generation = 10000
dictionary = {"fitness" : None,}

class HyperParameters:
    def __init__(self, num_inds, num_genes, tm_size, num_elites, num_parents, mutation_prob,
mutation_type):#collect parameters in a class
        self.num_inds = num_inds
        self.num_genes = num_genes
        self.tm_size = tm_size
        self.frac_elites = num_elites
        self.frac_parents = num_parents
        self.mutation_prob = mutation_prob
        self.mutation_type = mutation_type

class Gene:
    def __init__(self, index=-1, x=0, y=0, rad=1, R=0, G=0, B=0, A=0):
        self.index = index
        self.x = x
        self.y = y
        self.rad = rad
        self.R = R
        self.G = G
        self.B = B
        self.A = A

    def create_gene(self, index):#Create genes
        self.R = rnd.randrange(256)
        self.G = rnd.randrange(256)
```



```
self.B = rnd.randrange(256)
self.index = index
self.A = rnd.random()
temp_x = rnd.randrange(int(1.5 * WIDTH))
temp_y = rnd.randrange(int(1.5 * HEIGHT))
temp_rad = rnd.randrange(int(max(WIDTH, HEIGHT) / 2))
while not self.isIntersects(temp_x, temp_y, temp_rad):
    temp_x = rnd.randrange(int(1.5 * WIDTH))
    temp_y = rnd.randrange(int(1.5 * HEIGHT))
    temp_rad = rnd.randrange(int(max(WIDTH, HEIGHT) / 2))
self.x = temp_x
self.y = temp_y
self.rad = temp_rad

def isIntersects(self, x, y, r):#it checks whether genes are valid or not
    dist_x = abs(x - WIDTH / 2)
    dist_y = abs(y - HEIGHT / 2)

    if dist_x > (WIDTH / 2 + r): return False
    if dist_y > (HEIGHT / 2 + r): return False

    if dist_x <= (WIDTH / 2): return True
    if dist_y <= (HEIGHT / 2): return True

    cornerDistance_sq = (dist_x - WIDTH / 2) ** 2 + (dist_y - HEIGHT / 2) ** 2
    return cornerDistance_sq <= (r ** 2)

def guided_Mutation(self):#it provides guided_mutation according to information given
in manual
    #it actually does same things with create_gene() but there is some limitations
which is written
    temp_x = rnd.randrange(max(0, int(self.x-WIDTH/4)), int(self.x+WIDTH/4)+1)
    temp_y = rnd.randrange(max(0, int(self.y-HEIGHT/4)), int(self.y+HEIGHT/4)+1)
    temp_rad = rnd.randrange(max(0, self.rad-10), self.rad+11)
    while not self.isIntersects(temp_x, temp_y, temp_rad):
        temp_x = rnd.randrange(max(0, int(self.x-WIDTH/4)), int(self.x+WIDTH/4)+1)
        temp_y = rnd.randrange(max(0, int(self.y-HEIGHT/4)), int(self.y+HEIGHT/4)+1)
        temp_rad = rnd.randrange(max(0, self.rad-10), self.rad+11)
    self.x = temp_x
    self.y = temp_y
    self.rad = temp_rad
    self.R = rnd.randrange(max(0, self.R-64), min(self.R+65, 255))
    self.G = rnd.randrange(max(0, self.G-64), min(self.G+65, 255))
    self.B = rnd.randrange(max(0, self.B-64), min(self.B+65, 255))
    rnd_a = rnd.random()/2.0 - 0.25
    self.A = max(0, min(1.0, rnd_a + self.A))

def printGene(self):
```

```
    print(f"Gene - {self.index}: x:{self.x}, y:{self.y}, RAD:{self.rad}, R:{self.R},  
G:{self.G}, B:{self.B},A:{self.A}")  
  
class Individual:  
    def __init__(self,index = -1 ,fitness = 0,chro = []):  
        self.index = index  
        self.fitness = fitness  
        self.chro = chro  
  
    def create_Ind(self,index,num_of_genes):  
        self.chro = []  
        self.index = index  
        for i in range(num_of_genes):#create_genes and append to indivs  
            gene = Gene()  
            gene.create_gene(i+1)  
            self.chro.append(gene)  
  
    def created_image(self):  
  
        self.chro = sorted(self.chro, key=lambda ind: ind.rad, reverse=True)  
        white_image = np.zeros((HEIGHT,WIDTH,3),dtype=np.int8 )#firstly create white image  
and then make circle of each gene  
        cv2.rectangle(white_image, (0, 0), (WIDTH, HEIGHT), (255, 255, 255), -1)  
        for gene in self.chro:  
            overlay = white_image.copy()  
            cv2.circle(overlay,(gene.x,gene.y),gene.rad,(gene.B,gene.G,gene.R),-1)#make  
circles on overlayb of white image  
            white_image = cv2.addWeighted(overlay, gene.A, white_image, 1 - gene.A,  
0)#after that circles are added to white image  
            return white_image  
  
    def calculate_fitness(self):  
        #Firstly we should make order the chromosome  
        white_image = self.created_image()#firstly create white image with circle on it  
        diff = source_image.astype(float) - white_image.astype(float)#then calculate  
fitness with source_image and white_image  
        diff_squared = np.square(diff)  
        f = -np.sum(diff_squared)  
        self.fitness = f  
        return f  
  
    def printIndividual(self):  
        #print("Individual -",self.index)  
        print("Fitness: ", self.fitness)  
        #print("Chromosome:")  
        #for gene in self.chro:  
        #    gene.printGene()  
  
    def Mutation(self,mut_type,mut_prob):
```

```
        index_of_mut_genes = rnd.randrange(len(self.chro))#according to probability
        if mut_type == "unguided":#if it is unguided
            self.chro[index_of_mut_genes].create_gene(index_of_mut_genes)#then create new
gene without any restriction
        else:
            self.chro[index_of_mut_genes].guided_Mutation()#if it is guided then run guided
mutation
        mutation_points = np.random.randint(100,size=len(self.chro))
        mutation_points[index_of_mut_genes]=100
        for i in range (len(self.chro)):
            if mutation_points[i] < mut_prob*100:
                if mut_type == "unguided":#if it is unguided
                    self.chro[i].create_gene(i+1)#then create new gene without any
restriction
                else:
                    self.chro[i].guided_Mutation()#if it is guided then run guided mutation

    def Indv_cross(self,Indv2):
        crossover_points = np.random.randint(2,size=len(self.chro))#create an array that
takes random values which are 0,1. This array decides crossover genes
        chro1 = []#crossover genes
        chro2 = []
        for i in range(len(crossover_points)):
            if crossover_points[i] == 0:#if crossover_points is 0 then chro1 takes gene
from indv1
                chro1.append(self.chro[i])
                chro2.append(Indv2.chro[i])
            else:#if crossover_points is 0 then chro1 takes gene from indv2 and vice versa
                chro1.append(Indv2.chro[i])
                chro2.append(self.chro[i])
        child1 = Individual(index=self.index,chro=chro1)#then create indivs with crossovered
chromosomes
        child2 = Individual(index=Indv2.index,chro=chro2)
        return child1,child2

    def crossover_suggestion(self,parents):#suggestion 1 : choose best two from family
        children = []#save each childer in an array
        for i in range(0,len(parents),2):#run crossover between each double parents
            crossover_points =
np.random.randint(2,size=self.hyp_parameters.num_genes)#create an array that takes random
values which are 0,1. This array decides crossover genes
            chro1 = []#crossover genes
            chro2 = []
            family = []
            family.append(parents[i])
            family.append(parents[i+1])
            for j in range(0,len(crossover_points)-1):
                if crossover_points[j] == 0:#if crossover_points is 0 then chro1 takes gene
from indv1
```

```
        chro1.append(deepcopy(parents[i].chro[j]))
        chro2.append(deepcopy(parents[i+1].chro[j]))
    else:#if crossover_points is 0 then chro1 takes gene from indv2 and vice
versa
        chro1.append(deepcopy(parents[i+1].chro[j]))
        chro2.append(deepcopy(parents[i].chro[j]))
    child1 = Individual(index=parents[i].index,chro=chro1)#then create indvs with
crossovered chromosomes
    child2 = Individual(index=parents[i+1].index,chro=chro2)
    child1.calculate_fitness()
    child2.calculate_fitness()
    family.append(child1)
    family.append(child2)
    family.sort(key=lambda item: item.fitness, reverse=True)#sort the family
    children.append(family[0])#and choose the best two from four
    children.append(family[1])
    children = self.sortIndividuals(children) #we created children but do not forget
all children have index=1
    return children

class Population:
    def __init__(self,hyp_parameters,name,num_of_generations):
        self.hyp_parameters = hyp_parameters
        self.indvs = []
        self.name = name#I assign a name variable because it facilitates to giving name to
png files
        self.num_of_generations = num_of_generations
    def create_population(self):#create population according to given parameters
        self.indvs = []
        #self.best_indvs.create_Ind()
        for i in range(self.hyp_parameters.num_inds):
            indv = Individual()
            indv.create_Ind(i+1,self.hyp_parameters.num_genes)
            self.indvs.append(indv)

        # sort the population according to fitness value of indv
    def sortIndividuals(self,indvs):
        return sorted(indvs, key=lambda item: item.fitness, reverse=True)

    def printPopulation(self):
        for ind in self.indvs:
            ind.printIndividual()

    def eval_population(self):
        for indv in self.indvs:#calculate each one of the indvs in population
            indv.calculate_fitness()

    def selection(self):
        num_of_elits = int(self.hyp_parameters.frac_elites * self.hyp_parameters.num_inds)
```

```
        num_of_parents = int( self.hyp_parameters.frac_parents *
self.hyp_parameters.num_inds)
        #we need even number of parents
        if num_of_parents % 2 == 1:
            num_of_parents = num_of_parents + 1
        #sort the Individuals in the populations
        self.indvs = self.sortIndividuals(self.indvs)
        #now we can select the elits

        elits = deepcopy(self.indvs[:num_of_elits])
        others = deepcopy(self.indvs[num_of_elits:])
        parents = []
        #tournament between others
        for i in range(num_of_parents):
            best_one = rnd.randrange(len(others))
            for j in range(self.hyp_parameters.tm_size):
                index = rnd.randrange(len(others))
                if others[index].fitness > others[best_one].fitness:
                    best_one = index
            parents.append(others.pop(best_one))#best one does not contain in the others
because it is selected for parents now
        return(elits,parents,others)

    def selection_suggestion(self):
        num_of_elits = int(self.hyp_parameters.frac_elites * self.hyp_parameters.num_inds)
        num_of_parents = int( self.hyp_parameters.frac_parents *
self.hyp_parameters.num_inds)
        num_of_others = self.hyp_parameters.num_inds -num_of_elits-num_of_parents
        #we need even number of parents
        if num_of_parents % 2 == 1:
            num_of_parents = num_of_parents + 1
        #sort the Individuals in the populations
        self.indvs = self.sortIndividuals(self.indvs)
        #now we can select the elits

        elits = deepcopy(self.indvs[:num_of_elits])
        others = deepcopy(self.indvs[num_of_elits:])
        parents = []
        #tournament between others
        for i in range(num_of_parents):
            best_one = rnd.randrange(len(others))
            for j in range(self.hyp_parameters.tm_size):
                index = rnd.randrange(len(others))
                if others[index].fitness > others[best_one].fitness:
                    best_one = index
            parents.append(others.pop(best_one))#best one does not contain in the others
because it is selected for parents now
        new = Individual()
        new.create_Ind(others[-1].index,self.hyp_parameters.num_genes)
```

```
        new2 = Individual()
        new2.create_Ind(others[-2].index,self.hyp_parameters.num_genes)
        others.pop(len(others)-1)
        others.pop(len(others)-1)
        new.calculate_fitness()
        new2.calculate_fitness()
        others.append(new)
        others.append(new2)
        return(elits,parents,others)

def crossover(self,parents):
    children = []#save each childer in an array
    for i in range(0,len(parents),2):#run crossover between each double parents
        child1, child2 = parents[i].Indv_cross(parents[i+1])
        children.append(child1)
        children.append(child2)
        #children = self.sortIndividuals(children) #we created children but do not forget
all children have index=1
    return children

def crossover_sugg(self,parents):
    children = []#save each childer in an array
    for i in range(0,len(parents),2):#run crossover between each double parents
        child1, child2 = parents[i].crossover_suggestion(parents[i+1])
        children.append(child1)
        children.append(child2)
        #children = self.sortIndividuals(children) #we created children but do not forget
all children have index=1
    return children

def pop_mutation(self,population):
    for indv in population:
        if rnd.random() < self.hyp_parameters.mutation_prob:
            indv.Mutation(self.hyp_parameters.mutation_type,self.hyp_parameters.mutation_prob)#mutate
each individual
            self.eval_population()

def evolution(self):#we are going to make evolution to our population by specific
number of generation
    self.create_population()#create population
    fitness_values = []#record all fitness values and at the end of the funciton save
it by using json
    for i in range(self.num_of_generations):
        self.eval_population()#after creation of population evaluate the population
which means we calculate the fitness values of each member in the beginning
        elits,parents,others = self.selection()#then do selection and get elits,parents
and others
        children = self.crossover_sugg(parents=parents)#get children from parents
```



```
        self.pop_mutation(deepcopy(others+children))#skip the elits and make mutation
in rest of population
        self.indvs = deepcopy(others+children+elits)#now our populaion consists of
        others+children+elits
        fitness_values.append(elits[0].fitness)#lastly sort the individuals and save
the best one and do it every generations
        if i % 10 == 9:
            print(f>Loading %((i+1)/100),{fitness_values[i]}")
            #print(f"Children:{len(children)} , elits {len(elits)}, others
{len(others)} , parents {len(parents)}")
        if i % 1000 == 999:#pring png in each 1000 generations
            name = self.name+'iteration_'+str(i+1)+'.png'
            print("Name : "+name)
            cv2.imwrite(name,self.indvs[0].created_image())
    dictionary ={
        "fitness" : fitness_values
    }
    with open(self.name+"fitnes_values.json","w") as outfile:
        json.dump(dictionary,outfile)
    del fitness_values

def evolution2_2(self):
    self.create_population()#create population
    fitness_values = []#record all fitness values and at the end of the funciton save
it by using json
    for i in range(self.num_of_generations):
        self.eval_population()#after creation of population evaluate the population
which means we calculate the fitness values of each member in the beginning
        elits,parents,others = self.selection_suggestion()#then do selection and get
elits,parents and others
        children = self.crossover(parents=parents)#get children from parents
        self.pop_mutation(deepcopy(others+children))#skip the elits and make mutation
in rest of population
        self.indvs = deepcopy(others+children+elits)#now our populaion consists of
        others+children+elits
        fitness_values.append(elits[0].fitness)#lastly sort the individuals and save
the best one and do it every generations
        if i % 10 == 9:
            print(f>Loading %((i+1)/100),{fitness_values[i]}")
        if i % 1000 == 999:#pring png in each 1000 generations
            name = self.name+'iteration_'+str(i+1)+'.png'
            print("Name : "+name)
            cv2.imwrite(name,self.indvs[0].created_image())
    dictionary ={
        "fitness" : fitness_values
    }
    with open(self.name+"fitnes_values.json","w") as outfile:
        json.dump(dictionary,outfile)
    del fitness_values
```

```
def evolution2_2(self):
    self.create_population()#create population
    fitness_values = []#record all fitness values and at the end of the funciton save
it by using json
    for i in range(self.num_of_generations):
        self.eval_population()#after creation of population evaluate the population
which means we calculate the fitness values of each member in the beginning
        elits,parents,others = self.selection_suggestion()#then do selection and get
elits,parents and others
        children = self.crossover(parents=parents)#get children from parents
        self.pop_mutation(deepcopy(others+children))#skip the elits and make mutation
in rest of population
        self.indvs = deepcopy(others+children+elits)#now our populaiaon consists of
others+children+elits
        fitness_values.append(elits[0].fitness)#lastly sort the individuals and save
the best one and do it every generations
        if i % 10 == 9:
            print(f>Loading %{(i+1)/100},{fitness_values[i]}")
        if i % 1000 == 999:#pring png in each 1000 generations
            name = self.name+'iteration_'+str(i+1)+'.png'
            print("Name : "+name)
            cv2.imwrite(name,self.indvs[0].created_image())
    dictionary ={
        "fitness" : fitness_values
    }
    with open(self.name+"fitnes_values.json","w") as outfile:
        json.dump(dictionary,outfile)
    del fitness_values

def evolution2_3(self):#changable mut type and probability
    self.create_population()#create population
    fitness_values = []#record all fitness values and at the end of the funciton save
it by using json
    for i in range(self.num_of_generations):
        self.adjust_mut_parameters(i)
        self.eval_population()#after creation of population evaluate the population
which means we calculate the fitness values of each member in the beginning
        elits,parents,others = self.selection_suggestion()#then do selection and get
elits,parents and others
        children = self.crossover(parents=parents)#get children from parents
        self.pop_mutation(deepcopy(others+children))#skip the elits and make mutation
in rest of population
        self.indvs = deepcopy(others+children+elits)#now our populaiaon consists of
others+children+elits
        fitness_values.append(elits[0].fitness)#lastly sort the individuals and save
the best one and do it every generations
        if i % 10 == 9:
            print(f>Loading %{(i+1)/100},{fitness_values[i]}")
```

```
        if i % 1000 == 999: #pring png in each 1000 generations
            name = self.name+'iteration_'+str(i+1)+'.png'
            print("Name : "+name)
            cv2.imwrite(name,self.indvs[0].created_image())
    dictionary ={
        "fitness" : fitness_values
    }
    with open(self.name+"fitnes_values.json","w") as outfile:
        json.dump(dictionary,outfile)
    del fitness_values

def adjust_mut_parameters(self,iteration):
    if iteration < 1000:
        self.mutation_type = 'unguided'
        self.mutation_prob = 0.8
    elif iteration >5000:
        self.mutation_type = 'guided'
        self.mutation_prob = 0.5

"""
#suggestion 1 : choose best of two from family
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
popu(hyp_parameters=params,name="discussion/evol_2_1/",num_of_generations=num_generation)
pop.evolution2_1()

#suggestion 2 : get rid of the last two members of others(not elits and children) and
create new Individuals
params = HyperParameters(num_inds[4], num_genes[4], tm_size[3], frac_elites[2],
frac_parents[3], mutation_prob[3], mutation_type[1])
pop =
Population(hyp_parameters=params,name="discussion/evol_2_2/",num_of_generations=num_generat
ion)
pop.evolution2_2()
#?

#suggestion 3 adjustable mut type and probability
params = HyperParameters(num_inds[4], num_genes[4], tm_size[3], frac_elites[2],
frac_parents[3], mutation_prob[3], mutation_type[1])
pop =
Population(hyp_parameters=params,name="discussion/evol_2_2/",num_of_generations=num_generat
ion)
pop.evolution()
```

```
#default
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="num_inds/"+str(num_inds[2])+"/",num_of_generations=n
um_generation)
pop.evolution()

#inds
params = HyperParameters(num_inds[0], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="num_inds/"+str(num_inds[0])+"/",num_of_generations=n
um_generation)
pop.evolution()

params = HyperParameters(num_inds[1], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="num_inds/"+str(num_inds[1])+"/",num_of_generations=n
um_generation)
pop.evolution()

params = HyperParameters(num_inds[3], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="num_inds/"+str(num_inds[3])+"/",num_of_generations=n
um_generation)
pop.evolution()

params = HyperParameters(num_inds[4], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="num_inds/"+str(num_inds[4])+"/",num_of_generations=n
um_generation)
pop.evolution()

#genes
params = HyperParameters(num_inds[2], num_genes[0], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="num_genes/"+str(num_genes[0])+"/",num_of_generations
=num_generation)
pop.evolution()

params = HyperParameters(num_inds[2], num_genes[1], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[0])
```

```
pop =  
Population(hyp_parameters=params,name="num_genes/"+str(num_genes[1])+"/",num_of_generations  
=num_generation)  
pop.evolution()  
  
params = HyperParameters(num_inds[2], num_genes[3], tm_size[1], frac_elites[1],  
frac_parents[2], mutation_prob[1], mutation_type[0])  
pop =  
Population(hyp_parameters=params,name="num_genes/"+str(num_genes[3])+"/",num_of_generations  
=num_generation)  
pop.evolution()  
  
params = HyperParameters(num_inds[2], num_genes[4], tm_size[1], frac_elites[1],  
frac_parents[2], mutation_prob[1], mutation_type[0])  
pop =  
Population(hyp_parameters=params,name="num_genes/"+str(num_genes[4])+"/",num_of_generations  
=num_generation)  
pop.evolution()  
  
#tm_size  
params = HyperParameters(num_inds[2], num_genes[2], tm_size[0], frac_elites[1],  
frac_parents[2], mutation_prob[1], mutation_type[0])  
pop =  
Population(hyp_parameters=params,name="tm_size/"+str(tm_size[0])+"/",num_of_generations=num  
_generation)  
pop.evolution()  
  
params = HyperParameters(num_inds[2], num_genes[2], tm_size[2], frac_elites[1],  
frac_parents[2], mutation_prob[1], mutation_type[0])  
pop =  
Population(hyp_parameters=params,name="tm_size/"+str(tm_size[2])+"/",num_of_generations=num  
_generation)  
pop.evolution()  
  
params = HyperParameters(num_inds[2], num_genes[2], tm_size[3], frac_elites[1],  
frac_parents[2], mutation_prob[1], mutation_type[0])  
pop =  
Population(hyp_parameters=params,name="tm_size/"+str(tm_size[3])+"/",num_of_generations=num  
_generation)  
pop.evolution()  
  
#elits  
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[0],  
frac_parents[2], mutation_prob[1], mutation_type[0])  
pop =  
Population(hyp_parameters=params,name="elits/"+str(int(frac_elites[0]*100))+"/",num_of_gen  
erations=num_generation)  
pop.evolution()
```

```
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[2],
    frac_parents[2], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="elites/"+str(int(frac_elites[2]*100))+"/",num_of_gen
    erations=num_generation)
pop.evolution()

#parents
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
    frac_parents[0], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="parents/"+str(int(frac_parents[0]*100))+"/",num_of_g
    enerations=num_generation)
pop.evolution()

params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
    frac_parents[1], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="parents/"+str(int(frac_parents[1]*100))+"/",num_of_g
    enerations=num_generation)
pop.evolution()

params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
    frac_parents[3], mutation_prob[1], mutation_type[0])
pop =
Population(hyp_parameters=params,name="parents/"+str(int(frac_parents[3]*100))+"/",num_of_g
    enerations=num_generation)
pop.evolution()

#mut_prob
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
    frac_parents[2], mutation_prob[0], mutation_type[0])
pop =
Population(hyp_parameters=params,name="mut_prob/"+str(int(mutation_prob[0]*100))+"/",num_of
    _generations=num_generation)
pop.evolution()

params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
    frac_parents[2], mutation_prob[2], mutation_type[0])
pop =
Population(hyp_parameters=params,name="mut_prob/"+str(int(mutation_prob[2]*100))+"/",num_of
    _generations=num_generation)
```

```
pop.evolution()
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], 0.95, mutation_type[0])
pop =
Population(hyp_parameters=params,name="mut_prob/"+str(int(mutation_prob[3]*100))+"/",num_of
_generations=num_generation)
pop.evolution()

#type
params = HyperParameters(num_inds[2], num_genes[2], tm_size[1], frac_elites[1],
frac_parents[2], mutation_prob[1], mutation_type[1])
pop =
Population(hyp_parameters=params,name="mut_type/"+str(mutation_type[1])+"/",num_of_generati
ons=num_generation)
pop.evolution()
"""
```


Function of Print fitness:

```
import os
import matplotlib.pyplot as plt
import json
import numpy as np
x_list_9000 = []
x_list_10000 = []
for i in range(9000):#
    x_list_9000.append(i)
for i in range(10000):
    x_list_10000.append(i)
def plot_fitness_9000(file_name,folder):
    with open(file_name+'.json', 'r') as f:
        data = json.load(f)
        y = list(data.values())
        plt.figure()
        plt.plot(x_list_9000, y[0][1000:10000])
        # Customize the plot as desired
        plt.xlabel('Number of generation')
        plt.ylabel('Fitness Values')
        plt.title(folder+"/Iterations: 1000 - 10000")
        # Save the plot as a PNG file
        plt.savefig(file_name+'1000-10000.png')
        plt.close()
        f.close()
def plot_fitness_10000(file_name,folder):
    with open(file_name+'.json', 'r') as f:
        data = json.load(f)
        y = list(data.values())
        plt.figure()
        plt.plot(x_list_10000, y[0][0:10000])
        # Customize the plot as desired
        plt.xlabel('Number of generation')
        plt.ylabel('Fitness Values')
        plt.title(folder+"/Iterations: 1 - 10000")
        # Save the plot as a PNG file
        plt.savefig(file_name+'1-10000.png')
        plt.close()
        f.close()
folder_names =
["num_genes/", "num_inds/", "parents/", "tm_size/", "mut_type/", "mut_prob/", "elite
s/", "discussion/"]
for i in folder_names:
    folders = [f for f in os.listdir(i) if os.path.isdir(os.path.join(i, f))]
    for folder in folders:
        plot_fitness_10000(str(i)+str(folder)+"/fitnes_values",str(i)+str(folder))
        plot_fitness_9000(str(i)+str(folder)+"/fitnes_values",str(i)+str(folder))
        #enter each folder and find all folders in each defined folder
```

Print nine images in one image:

```
import os
import matplotlib.pyplot as plt
import numpy as np
import cv2

def plot_images(file_names, folder_name):
    # Generate some random images for demonstration
    images = [cv2.cvtColor(cv2.imread(file_name), cv2.COLOR_BGR2RGB) for
file_name in file_names]
    # Create a 3x3 grid of subplots
    fig, axs = plt.subplots(3, 3)

    # Iterate over the axes and images, and plot each image on a subplot
    for ax, img in zip(axs.flat, images):
        ax.imshow(img)
        ax.axis('off')

    # Set the title of the figure
    fig.suptitle(folder_name+"/Iterations: 2000 - 10000 ")

    # Show the plot
    plt.savefig(folder_name+"/nine_images.png")
    plt.close()

files= []
folder_names =
["num_genes/", "num_inds/", "parents/", "tm_size/", "mut_type/", "mut_prob/", "elite
s/", "discussion/"]
for i in folder_names: #enter each folder and find all folders in each defined
folder

    folders = [f for f in os.listdir(i) if os.path.isdir(os.path.join(i, f))]
    for folder in folders:
        for count in range(2000,11000,1000):
            folder_name = str(i)+str(folder)
            files.append(folder_name+"/iteration_"+str(count)+'.png')
        plot_images(files, folder_name)
        files=[]
```