Parameter settings

```
gene_length = 30 #二進制的長度
population_size= 500 #每一代染色體的數量
mutation_rate = 0.5 #變異機率
tolerance = 1e-11 # 停止條件
aus = 1e6
target = float(input("請輸入您要計算的根號值:"))
```

def Binary(chromosome)

這個部分主要是將我的數值轉為二進制的定義

def decimal_number(chromosome)

這個部分主要是將我的數值轉為十進制的定義

def generate_population(population_size, gene_length)

生成初始族群的函數

- def fitness_function(individual)
 - # 評估染色體適應度的函數

```
chromosome = decimal_number(individual) / (10**11)
score = aus - abs(chromosome**2 - target)
return max(0, score) #確保適應度值大於等於零
```

def crossover(parent1, parent2)

交叉遺傳的函數

```
# 隨機選擇交叉點,避免在邊界處交叉
point = random.randint(1, len(parent1) - 2)
# 生成子代 1, 將父母 1 的前段和父母 2 的後段結合
child1 = parent1[:point] + parent2[point:]
# 生成子代 2, 將父母 2 的前段和父母 1 的後段結合
child2 = parent2[:point] + parent1[point:]
return child1, child2
```

def mutate(individual)

變異的函數

隨機選擇染色體中的一個位置
mutated_index = random.randint(0, len(individual) - 1)
將選定位置的基因取反,即 0 變為 1 , 1 變為 0
mutated_gene = '0' if individual[mutated_index] == '1' else '1'
產生變異後的染色體,將選定位置的基因替換為變異後的基因
return individual[:mutated_index] + mutated_gene +

individual[mutated_index + 1:]

def genetic_algorithm()

遺傳演算法主要的函數,會先生成一個族群,然後將族群按照適應的程度 去排序,產生最佳的染色體,這邊也會去選擇父母的染色體,進行交叉遺傳,然後變異,將好的子# 檢查是否達到停止條件,子代留下來,產生新的族群

(以下為執行結果)



(圖一) 可以輸入需要計算的根號數字



(圖二)為輸入(根號)5所得到最終的結果

(以下為程式碼)

```
import random
gene_length = 50 #binary length
population_size= 1000 #The number of chromosomes in each generation
mutation_rate = 0.4 #Mutation probability
tolerance = 1e-11 # Stop condition
def Binary(chromosome):
    #Convert to binary(str)
    binary_string=bin(chromosome)[2:]
    return binary_string
        #轉乘10進制(str)
return int(chromosome,2)
def generate_population(population_size,gene_length):
    return[''.join(random.choice('01') for _ in range(gene_length)) for _ in range(population_size)]
def fitness_function(individual):
    chromosome = decimal_number(individual) / (10**11)
    score = aus - abs(chromosome*2 - target)
    return max(0, score) # 誘保适应度值大于等于零
def select_parents(population):
#選擇2個染色體(當父母)
        #population為這一代中所有的父母
return random.choices(population, k=2, weights=[fitness_function(ind) for ind in population])
        crossover(parent), parent/):
#(1, len(parent1) - 2)是為了避免運動投票局
point = random.randint(1, len(parent1) - 2)
child1 = parent1[:point] + parent2[point:]
child2 = parent2[:point] + parent1[point:]
return child1, child2
def mutate(individual):
#選擇一個點進行變異
#使1-->0 和 0-->1
        mutated_index = random.randint(0, len(individual) - 1)
mutated_index = '0' if individual[mutated_index] == '1' else '1'
return individual[:mutated_index] + mutated_gene + individual[mutated_index+1:]
def genetic_algorithm():
    #Randomly generate the first generation chromosomes
population=generate_population(population_size,gene_length)
generation=0
             hile True:
generation+=1
#Mank dyeing replacements with high scores ranked first
population=sorted(population,key=fitness_function,reverse=True)
new_population=[population[0]]
#Generate the next generation (population_size chromosomes)
while len(new_population)>population_size:
#Choose parents to have children
parent1,parent2=select_parents(population)
child1,child2=crossover(parent1,parent2)
                       if random.random()<mutation_rate:
    child2=mutate(child2)
                 #The best solution of this generation
ans=population[0]
a=fitness_function(ans)
                print('祝所有的AIIA學長都可以準時畢業離校')
         print("他們想時紅山學校節的以來明確無應從 )
print(f*Root (int(target)) :', decimal_number(ans) / 10**11)
print(f'Closest target number {int(target)} :', (decimal_number(ans) / 10**11)**2)
        __name__ == '__main_
genetic_algorithm()
```