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import numpy as np
import matplotlib.pyplot as plt
# Datele de intrare
population_size = 20
                              # dimensiunea populatiei
#lowerBound = None
                                # limita de jos a domeniului
#upperBound = None
                                # limita de sus
domain = (-1, 2)
                            # domeniul functiei
coefficients = (-1, 1, 2)
                            # parametrii functiei
                           # precizia de discretizare a intervalului
precision = 6
crossover_probability = 0.25
                                 # probabilitatea de recombinare
mutatioin_probability = 0.01
                                # probabilitatea de mutatie
generations = 50
                             # numarul de generatii
leap = 25
                         # numarul de generatii peste care se sare la plotare
f = open("Evolutie.txt", "w")
# Calculam numarul de biti necesari pentru precizia si domeniul date
domain range = domain[1] - domain[0]
total values = domain_range * (10 ** precision)
bits needed = int(np.ceil(np.log2(total values)))
#print(bits_needed)
# Functia de fitness
def fitness(x):
  return coefficients[0] * x ** 2 + coefficients[1] * x + coefficients[2]
def binary_to_double(bit_string):
  max val = 2 ** len(bit string) - 1
  int val = int(bit string, 2)
  real_val = domain[0] + (domain_range * int_val) / max_val
  return real val
def binary search(array, target):
  left = 0
  right = len(array) - 1
  while left <= right:
     mid = (left + right) // 2
     if array[mid] < target:
       left = mid + 1
     else:
       right = mid - 1
  return left
def selection binary search(population, fitness vals):
  fitness sum = sum(fitness vals)
  probabilities = [f / fitness_sum for f in fitness_vals]
  cumulative prob = np.cumsum(probabilities)
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indices_prob = []
  for in range(population size - 1):
     r = np.random.rand()
     index = binary search(cumulative prob, r)
     selected indices.append(index)
     indices prob.append(r)
  selected = [population[i] for i in selected indices]
  selected indices.append(np.argmax(fitness vals))
  best individual = population[np.argmax(fitness vals)]
  selected.append(best individual)
  return (selected, selected_indices, indices_prob)
# Functii pentru afisare
def print best mean fitness(real vals, fitness vals):
  best fitness = fitness vals[np.argmax(fitness vals)]
  best val = real vals[np.argmax(fitness vals)]
  mean_fitness = sum(fitness_vals) / population_size
  f.write(f"x = {(best val + 1):.6f} Best fitness: {best fitness} Mean fitness: {mean fitness}\n")
def print_initial_population(population, real_vals, fitness_vals):
  f.write("Populatia Initiala: \n")
  for i, (bit string, real val, fitness) in enumerate(zip(population, real vals, fitness vals)):
     f.write(f" \{i + 1\}: \{bit string\} x = \{real val: .6f\} f = \{fitness\}")
     f.write("\n")
def print selection prob(fitness vals):
  fitness_sum = sum(fitness_vals)
  probabilities = [f / fitness sum for f in fitness vals]
  cumulative_prob = np.cumsum(probabilities)
  f.write("Probabilitati Selectie: \n")
  for cromozome in range(population_size):
     f.write(f"Cromozom {cromozome + 1}: probabilitate: {probabilities[cromozome]}\n")
  f.write("Intervale Probabilitati Selectie: \n")
  for item in cumulative_prob:
     f.write(str(item))
     f.write("\n")
def print_selections(selected, selected_prob):
  for i in range(len(selected) - 1):
     f.write(f"u = {selected prob[i]} selectam cromozomul {selected[i] + 1\\n")
  f.write(f"u = Elitism
                          selectam cromozomul {selected[i + 1] + 1}\n")
```

selected indices = []

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def print selected cromozomes(selected, selected ind, real vals, fitness):
  f.write("Dupa selectie: \n")
  for i in range(len(selected)):
     f.write(f"{i + 1}: {selected[i]} x = {real\_vals[selected\_ind[i]]:.6f} f = {fitness[selected\_ind[i]]}\n"
def print pair crossover prob(index1, chromosome1, index2, chromosome2, probability):
  if(index1 == 0):
     f.write(f"Probabilitatea de incrucisare este: {crossover_probability}\n")
  f.write(f"{index1 + 1}: {chromosome1} ")
  f.write(f"{index2 + 1}: {chromosome2}\n")
  if(probability < crossover probability):
     f.write(f"{probability} < {crossover_probability} -> participa\n")
def print_crossover_result(index1, chromosome1, index2, chromosome2, crossover_point, offspring1,
offspring2):
  f.write(f"Recombinare dintre cromozomul {index1 + 1} cu cromozomul {index2 + 1}:\n")
  f.write(f"{chromosome1} {chromosome2} punct {crossover point}\n")
  f.write(f"Rezultat {offspring1} {offspring2}\n")
def print_crossover_offsprings(offsprings):
  for i in range(len(offsprings)):
     real_val = binary_to_double(offsprings[i])
     f.write(f"{i + 1}: {offsprings[i]} ")
     f.write(f"x = {real\_val:.6f} ")
     f.write(f"f = {fitness(real_val)}\n")
def print_mutated_indices(mutated_indices):
  f.write(f"Probabilitatea de mutatie pentru fiecare gena {mutatioin probability}\n")
  f.write("Au fost modificati cromozomii: \n")
  for index in mutated indices:
     f.write(f"{index + 1}\n")
def print_mutation_offsprings(offsprings):
  f.write("Dupa mutatie: \n")
  for i in range(len(offsprings)):
     real_val = binary_to_double(offsprings[i])
     f.write(f"{i + 1}: {offsprings[i]} x = \{real\_val:.6f\} f = \{fitness(real\_val)\}\n"
def print_generation_visual(real_vals, fitness_vals, generation):
  real_vals_distrib = np.linspace(domain[0], domain[1], 150)
  fitness_vals_distrib = fitness(real_vals_distrib)
  plt.scatter(real_vals, fitness_vals, color = "blue")
  plt.plot(real vals distrib, fitness vals distrib)
  plt.xlabel("values")
  plt.ylabel("fitness")
  plt.title(f"Generation: {generation + 1}")
  plt.show()
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```
def Crossover(population size, chromosomes, generation):
  offspring = []
  for i in range(0, population size, 2):
     parent1 = chromosomes[i]
     parent2 = chromosomes[min(i + 1, len(chromosomes) - 1)]
     u = np.random.rand()
     if generation == 0:
       print pair crossover prob(i, parent1, min(i + 1, len(chromosomes) - 1), parent2, u)
     if u < crossover probability:
       crossover_point = np.random.randint(1, bits_needed - 1)
       offspring1 = parent1[: crossover point] + parent2[crossover point:]
       offspring2 = parent2[ : crossover_point] + parent1[crossover_point : ]
       if generation == 0:
          print_crossover_result(i, parent1, min(i + 1, len(chromosomes) - 1), parent2, crossover_point,
                       offspring1, offspring2)
     else:
       offspring1, offspring2 = parent1, parent2
     offspring.extend([offspring1, offspring2])
  return offspring
def print_init_crossover_prob(chromosomes, participant_indices, participant_prob):
  f.write(f"Probabilitatea de incrucisare este {crossover_probability}\n")
  counter = 0
  for i in range(len(chromosomes)):
     #f.write(f"{i + 1}: {chromosomes[i]} u = {participant_prob[i]}")
     if i in participant indices:
       f.write(f"\{i + 1\}: \{chromosomes[i]\}\ u = \{participant\_prob[i]\} < \{crossover\_probability\}\ participa<math>\n")
     else:
       f.write(f"{i + 1}: {chromosomes[i]} u = {participant_prob[i]}\n")
def Crossover_2(population_size, chromosomes, generation):
  participant_indices = []
  participant_probabilities = []
  for i in range(population_size):
     u = None
     if(i == population_size - 1):
       u = 2
     else:
       u = np.random.rand()
     participant probabilities.append(u)
     if u < crossover_probability:
       participant_indices.append(i)
  if generation == 0:
     print_init_crossover_prob(chromosomes, participant_indices, participant_probabilities)
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for i in range(population_size):
     if i in participant indices:
       participant_indices.remove(i)
       for j in range(i, population size):
          if j in participant indices:
             parent1 = chromosomes[i]
             parent2 = chromosomes[j]
             crossover_point = np.random.randint(1, bits_needed - 1)
             offspring1 = parent1[: crossover point] + parent2[crossover point:]
             offspring2 = parent2[: crossover point] + parent1[crossover point:]
             chromosomes[i] = offspring1
             chromosomes[i] = offspring2
             participant_indices.remove(j)
             if generation == 0:
               print_crossover_result(i, parent1, j, parent2, crossover_point,
                       offspring1, offspring2)
             break
  return chromosomes
def Mutate(offsprings):
  mutated_offspring = []
  mutated_indices = []
  for i in range(len(offsprings)):
     mutated individual = None
     if(i == len(offsprings) - 1):
       mutated individual = offsprings[i]
     else:
       mutated individual = ".join(
          bit if np.random.rand() > mutatioin_probability else '1' if bit == '0' else '0'
          for bit in offsprings[i]
     mutated_offspring.append(mutated_individual)
     if mutated_individual != offsprings[i]:
       mutated_indices.append(i)
  return (mutated_offspring, mutated_indices)
# Generare populatie initiala
population = [".join(np.random.choice(['0', '1']) for _ in range(bits_needed)) for _ in
range(population_size)]
real_vals = [binary_to_double(individual) for individual in population]
fitness_vals = [fitness(x) for x in real_vals]
print_initial_population(population, real_vals, fitness_vals)
print_selection_prob(fitness_vals)
print_generation_visual(real_vals, fitness_vals, 0)
for generation in range(generations):
  # Evaluarea fitness-ului
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real vals = [binary to double(individual) for individual in population]
  fitness\_vals = [fitness(x) for x in real\_vals]
  if (generation + 1) % leap == 0:
     print generation visual(real vals, fitness vals, generation)
  if generation != 0:
     print best mean fitness(real vals, fitness vals)
  # Selectia
  selected, selected ind, ind prob = selection binary search(population, fitness vals)
  if generation == 0:
     print selections(selected ind, ind prob)
     print_selected_cromozomes(selected, selected_ind, real_vals, fitness_vals)
  # Crossover
  offsprings = Crossover_2(population_size, selected, generation)
  if generation == 0:
     print_crossover_offsprings(offsprings)
  # Mutatii
  mutated_offsprings, mutated_indices = Mutate(offsprings)
  if generation == 0:
     print_mutated_indices(mutated_indices)
     print mutation offsprings(mutated offsprings)
  # Updatam populatia pentru urmatoarea generatie
  population = mutated offsprings
final_real_vals = [binary_to_double(individual) for individual in population]
final_fitness_vals = [fitness(x) for x in final_real_vals]
best_sol_index = np.argmax(final_fitness_vals)
best sol real val = final real vals[best sol index]
best_sol_fitness = final_fitness_vals[best_sol_index]
print(best_sol_real_val, best_sol_fitness)
```