

### 1 Exercise PP. 7 - 1

Considering the Rosenbrock's parabolic valley:

$$f(x_1, x_2) = 100(x_1 x_1^2)^2 + (1 - x_1)^2$$

Obtain the point of minima using a Genetic Algorithm.

### 1.1 MATLAB Files

The genetic algorithm that follows performs the minimisation of the objective function for 10 different initial populations and outputs the results from the first one and the average of all of them, plotting the search path, the population at different generation and a comparison graph between the standard deviation, the mean and the best fitness values.

#### GA\_main.m

```
%% GA_main.m
   % Main script of the GA
3
   % Calls: genetic
4
5
   addpath([pwd,'\Functions';],[pwd,'\Functions\Plot'],[pwd,'\
      Functions\Genetic_Operators']);
6
   clc
   clear
8
   close all
9
10
   time_total = 0;
11
   X_{total} = [0,0];
12
   f_{total} = 0;
13
  fitness_total = 0;
14
   k_{total} = 0;
15
16
   for m = 1:10
17
       [time, X_best, best_fitness,f,k] = genetic(m);
18
       time_total = time_total + time;
19
       X_total = X_total + X_best;
20
       fitness_total = fitness_total + best_fitness(1);
21
       k_total = k_total + k;
22
23
   end
24
25
   time_average = time_total/m;
26
   X_average = X_total/m;
   fitness_average = fitness_total/m;
```



```
28
  |k_average = k_total/m;
29
30
  % Results Print
31
32 | f_average = f(X_average(1), X_average(2));
   error = abs(f_average-f(1,1));
34
   accuray = (1-error)*100;
35
   fprintf([ ...
36
37
             'Average results of %i different initial populations:
                \n', ...
38
             'Number of generations: %.0f\n', ...
39
            'Approximate Solution: [%.5f , %.5f]\n', ...
40
            'Objective function value: %.4f \n', ...
            'Maximum fitness value: %.4f \n', ...
41
            'Precision: %.5f%% \n', ...
42
43
            'Average time: %.4fs\n'], ...
            m, k_average, X_average(1), X_average(2), f_average,
44
               fitness_average, accuray, time_average)
```

#### PP7 data.m

```
%% PP7_data.m
   % Problem Data
3 % Calls: truss_constraints
4
  % Data
5
6 | 1b = [0; 0]; % Lower limits
7
   ub = [2; 2]; % Upper limits
9
   % Objective Function
10 | f = 0(x1, x2) 100.*(x2-x1.^2).^2+(1-x1).^2;
11
12 | % Population Parameters
13 \mid N_{pop} = 100;
                     % Number of individuals
14 \mid 1_c = 300;
                       % Chromosome Length
15 | b = [1_c/2 1_c/2]; % Number of alleles per gene
16
17
  % Max number of iterations
18 \mid kmax = 1000;
19
20 | % Tolerance
  tol = 1e-6;
```



### genetic.m

```
%% genetic.m
  % Genetic algorithm
  % Calls: truss_data, population_generation, fitness_function,
      crossover, mutation, Evolutionary_cycle, plot_bestfitness,
      plot_bestfitness_path, plot_pop_evolution,
      plot_pop_evolution2
4
   function [time, X_best, fitness, f, k] = genetic(m)
5
6
   tic
  % Data Initialization
9
  PP7_data
10
11 % Initial Population of solutions generation
12 | [Population] = population_generation(N_pop,l_c);
13 k=0;
14
15 \% Fitness Evaluation
16 | pop_final=0;
17
  [fitness, X_pop, Population] = fitness_function(Population, N_pop,
      lb,ub,l_c,f);
18
19
   if m == 1
20
       % Plot - Fittest Individual for each Generation
21
       plot_bestfitness(f,lb,ub);
22
   end
23
24
   while k<kmax && std(fitness)>tol
25
26
       k=k+1;
27
28
       % Save Current Generation
29
       X_{pop_old} = X_{pop};
30
31
       if m == 1
32
            pop_evolution(:,:,k) = X_pop;
33
            std_fitness(k) = std(fitness);
34
           mean_fitness(k) = mean(fitness);
35
            best_fitness(k) = fitness(1);
36
       end
       % Genetic Operators
38
```



```
39
       Offsprings_population = crossover(l_c,N_pop,fitness,
          Population);
40
       Offsprings_population = mutation(Offsprings_population, size
          (Offsprings_population,1),l_c);
41
42
       % Fitness evaluation of the individuals in the sons'
          population
43
       [fitness_offsprings,~,Offsprings_population] =
          fitness_function(Offsprings_population, size(
          Offsprings_population,1),lb,ub,l_c,f);
44
45
       [fitness, ~, Population] = fitness_function(Population, N_pop,
          lb,ub,l_c,f);
46
47
       % Survivors selection
       [Population_new,fitness_new] = Evolutionary_cycle(
48
          Population, N_pop, Offsprings_population, fitness,
          fitness_offsprings);
49
       Population = Population_new;
       fitness = fitness_new;
51
52
       % Decoding
53
       X_pop = decoding(Population, N_pop, lb, ub, l_c);
54
       if m == 1
55
            % Plot - Current Generation
56
57
           plot_bestfitness_path(X_pop, X_pop_old, pop_final)
58
       end
59
60 end
61
   % Final Population Fittest Individual
62
63
   X_{best} = X_{pop}(1,:);
64
  time = toc;
65
66
67
   if m == 1
68
       % Plot - Final Population
69
       pop_final=1;
70
       plot_bestfitness_path(X_pop, X_pop_old, pop_final);
71
72
       % Plot - Generations Evolution
73
       pop_evolution(:,:,k+1) = X_pop;
74
       plot_pop_evolution(pop_evolution, X_pop, k, f, lb, ub);
```



```
76
       % Plot - Generations Evolution - Standard Deviation and
       std_fitness(k+1) = std(fitness);
77
       mean_fitness(k+1) = mean(fitness);
78
       best_fitness(k+1) = fitness(1);
79
80
       plot_pop_evolution2(std_fitness, mean_fitness, best_fitness, k
          );
81
82
       % Results
83
       fobj=f(X_best(1), X_best(2));
84
       fprintf(['First Initial Population: \n',...
85
                 'Number of generations: %d\n', ...
                 'Approximate Solution: [%.5f , %.5f]\n', ...
86
87
                 'Objective function value: %.4f \n\n\n'], ...
88
                 k, X_best(1), X_best(2), fobj)
89
   end
```

### population\_generation.m

```
%% population_generation.m
  % Initial population generation
2
3
4
  function [Population] = population_generation(N_pop,l_c)
5
  Population1 = randi([0 1], N_pop, l_c);
6
8
  % Eliminate Clones
  Population = unique(Population1, 'rows');
10
11
  while size(Population1,1) > size(Population,1)
12
       difference = size(Population1,1) - size(Population,1);
       Population_add = randi([0 1], difference, l_c);
13
       Population1 = [Population; Population_add];
14
15
       Population = unique(Population1, 'rows');
16
   end
```

### Evolutionary\_cycle.m



#### fittness\_fucntion.m

```
%% fitness_function.m
  % Fitness Function
  % Calls: decoding
4
   function [fitness, X_real_ordered, population_ordered] =
5
      fitness_function(population, N_pop, lb, ub, l_c, f)
6
7
  % Decoding
8
   X_real = decoding(population, N_pop, lb, ub, l_c);
9
10 % Fitness evaluation
11 | fitness=-f(X_real(:,1),X_real(:,2));
12
13 | % Sorted Population (descend)
14 | [fitness, position_pop] = sort(fitness, 'descend');
15 | X_real_ordered = X_real(position_pop,:);
16
  population_ordered = population(position_pop,:);
17
  end
```

### decoding.m



```
X_real=zeros(N_pop,2);
9
10
   for i=1:N_pop
11
        for j=1:1_c/2
12
            %x1
13
            X_{bin}(i,1) = X_{bin}(i,1) + Population(i,j) *2^(1_c/2-j);
14
15
16
            X_{bin}(i,2) = X_{bin}(i,2) + Population(i,j+1_c/2) *2^(1_c/2-
                j);
17
        end
18
   end
19
20
   for i=1:N_pop
21
        %x1
22
        X_{real}(i,1) = lb(1)+(ub(1)-lb(1))/(2^(l_c/2) - 1)*X_bin(i)
           ,1);
23
24
        %x2
25
        X_{real}(i,2) = lb(2) + (ub(2) - lb(2)) / (2^{(1_c/2)} - 1) * X_bin(i)
           ,2);
26
   end
```

#### selection.m

```
%% selection.m
2
   % Selection Operator
3
4
   function [parent1, parent2, roulette] = selection(fitness, N_pop
      ,population,i, roulette)
5
   if i == 1
6
7
       % Selection probability for each individual
       selection_prob = fitness/sum(fitness);
8
9
       % Roulette intervals
10
       roulette=zeros(N_pop,1);
11
12
       roulette(1) = selection_prob(1);
13
       for i=2:N_pop
14
           roulette(i)=roulette(i-1) + selection_prob(i);
15
       end
16
   end
17
18 | % Select first parent
```



```
selection_value = rand();
parent_position1 = find(selection_value < roulette,1);
parent1 = population(parent_position1,:);

% Select second parent
selection_value = rand();
parent_position2 = find(selection_value < roulette,1);

while parent_position2 == parent_position1
selection_value = rand();
parent_position2 = find(selection_value < roulette,1);
end

parent2 = population(parent_position2,:);</pre>
```

#### crossover.m

```
%% crossover.m
  % Crossover operator
  % Calls: selection
4
5
   function [offspring_pop] = crossover(l_c,N_pop,fitness,
      population)
6
   offspring_pop = zeros(N_pop,l_c);
8 \mid son1 = zeros(1,l_c);
9 \mid son2 = zeros(1,l_c);
10 | roulette=0;
11
12 | for i=1: N_pop/2
13
       % Parent Selection
14
        [parent1, parent2, roulette] = selection(fitness, N_pop,
          population, i, roulette);
15
16
       % Mask
17
       mask = randi([0,1],1,l_c);
18
19
       for j=1:1_c
20
            if mask(j) == 0
21
                son1(j) = parent1(j);
22
                son2(j) = parent2(j);
23
            else
24
                son1(j) = parent2(j);
25
                son2(j) = parent1(j);
```



#### mutation.m

```
%% mutation.m
2
   % Mutation operator
   function [offspring_pop] = mutation(offspring_pop, N_pop, l_c)
4
5
6
   for i=1:N_pop
7
       for j=1:1_c
8
            prob_mut = rand();
9
10
            if prob_mut <= 0.01</pre>
11
                 if offspring_pop(i,j) == 0
12
                     offspring_pop(i,j) = 1;
13
                 else
14
                     offspring_pop(i,j) = 0;
15
                 end
16
            end
17
       end
18
   end
```

### plot\_bestfitness.m

```
%% File: plot_bestfitness.m
  % Plot fittest indivual for each generation
3
  function [fig] = plot_bestfitness(ff,lb,ub)
4
5
  fig = figure(1);
6
7 | fcontour(ff,[lb(1) ub(1) lb(2) ub(2)])
  title('$f(x_1,x_2)=100(x_2-x_1^2)^2+(1-x_1)^2$','Interpreter','
     latex')
  xlabel('$x_1$',Interpreter='latex')
  ylabel('$x_2$',Interpreter='latex')
10
  grid on
11
  hold on
12
```



### plot\_bestfitness \_path.m

```
%% plot_bestfitness_path.m
  % Plot - Current Population
3
4
  function [fig] = plot_bestfitness_path(X_pop, X_pop_old,
     pop_final)
5
6
  if pop_final == 0
       plot([X_pop_old(1,1), X_pop(1,1)],[X_pop_old(1,2),X_pop(1,2)
         ],'o-r');
8
  else
9
       plot([X_pop_old(1,1),X_pop(1,1)],[X_pop_old(1,2),X_pop(1,2)
          ], 'o-k', MarkerFaceColor='k');
  end
11
  drawnow
```

### plot\_pop \_evolution.m

```
%% plot_pop_evolution.m
   % Generation evolution plot
3
4
   function [fig] = plot_pop_evolution(pop_evolution, X_pop,k,ff,lb
      ,ub)
5
6
   fig = figure(2);
   tiledlayout('flow')
   a=round((k+1)/5);
9
  for i=1:a:(1+4*a)
10
11
       nexttile
12
       fcontour(ff,[lb(1) ub(1) lb(2) ub(2)])
13
       xlabel('$x_1$','Interpreter','latex')
       ylabel('$x_2$','Interpreter','latex')
14
       hold on
15
16
17
       scatter(pop_evolution(:,1,i),pop_evolution(:,2,i),"red")
18
       title(['Generation ',num2str(i-1)],'Interpreter','latex')
19
   end
20
21 nexttile
22 | fcontour(ff,[lb(1) ub(1) lb(2) ub(2)])
23 | xlabel('$x_1$','Interpreter','latex')
24 | ylabel('$x_2$','Interpreter','latex')
```



```
hold on

scatter(X_pop(:,1), X_pop(:,2), "red")

title(['Generation ',num2str(k)], 'Interpreter', 'latex')
```

### plot\_pop \_evolution2.m

```
%% plot_pop_evolution2.m
   % Generation Evolution plot - stantdard deviation and mean
3
4
   function [fig] = plot_pop_evolution2(std_fitness, mean_fitness,
      best_fitness,k)
6
   fig = figure(3);
   plot(0:k,std_fitness,"blue",'LineWidth',2) % Plot standard
      deviation
9
   hold on
10
   plot(0:k,mean_fitness,"red",'LineWidth',2) % Plot mean
11
12 hold on
13
14
   plot(0:k,best_fitness, "green", 'LineWidth', 2) % Plot best
      fitness
15
16 | title('Fitness', 'Interpreter', 'latex')
   xlabel('Generation','Interpreter','latex')
17
   xlim([0 k])
18
19
20
   legend('Standard Deviation', 'Mean', 'Best Fitness', 'Interpreter'
           'latex', 'Location', 'northeast')
21
```



### 1.2 Results

The results below were obtained for a population with 100 individuals, a chromosome length equal to 300 and with 150 alleles per gene. The precision shown in the results was calculated using the objective function value for the exact point of minima, which is f(1,1) = 0.

First Initial Population: Number of generations: 152

Approximate Solution: [0.99992, 0.99985]

Objective function value: 0.0000

Average results of 10 different initial populations:

Number of generations: 142

Approximate Solution: [0.99989, 0.99978]

Objective function value: 0.0000 Maximum fitness value: -0.0000

Precision: 100.00000% Average time: 1.1266s





