

1 Exercise PP. 8

Consider the following optimisation problem:

$$\begin{aligned} &\text{minimise: } f(x_1, x_2) = 2(x_1 - 2)^2 + 4(x_2 - 1)^2 \\ &\text{subject to: } 2x_1 + 8x_2 \leq 6 \\ &\quad \quad \quad 2x_1 \geq 2x_2 \end{aligned}$$

Solve the constrained minimisation 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

```
1 %% GA_main.m
2 % Main script of the GA
3 % Calls: genetic
4
5 addpath([pwd, '\Functions'];),[pwd, '\Functions\Plot'],[pwd, '\
    Functions\Genetic_Operators']);
6 clc
7 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;
27 fitness_average = fitness_total/m;
28 k_average = k_total/m;
29
30
31 % Results Print
32 f_average = f(X_average(1),X_average(2));
33 error = abs((f_average-f(5/3,1/3))/f(5/3,1/3));
34 accuray = (1-error)*100;
35
36 fprintf([ ...
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', ...
41         'Maximum fitness value: %.4f \n', ...
42         'Precision: %.5f%% \n', ...
43         'Average time: %.4fs\n'], ...
44         m, k_average, X_average(1),X_average(2),f_average,
          fitness_average, accuray, time_average)
```

PP8_data.m

```
1 %% PP8_data.m
2 % Problem Data
3 % Calls: truss_constraints
4
5 % Data
6 lb = [0; 0]; % Lower limits
7 ub = [2; 0.5]; % Upper limits
8
9 % Objective Function
10 f = @(x1,x2) 2.*(x1-2).^2 + 4.*(x2-1).^2;
11
12 % Constraints
13 syms x1 x2
14 c(1,1) = 2*x1+8*x2-6;
15 c(1,2) = -2*x1+2*x2;
16 g = matlabFunction(c);
17
18 % Population Parameters
19 N_pop = 100; % Number of individuals
20 l_c = 300; % Chromossome Length
```

```
21 b = [l_c/2 l_c/2]; % Number of alleles per gene
22
23 % Max number of iterations
24 kmax = 1000;
25
26 % Tolerance
27 tol = 1e-6;
```

genetic.m

```
1 %% genetic.m
2 % Genetic algorithm
3 % Calls: truss_data, population_generation, fitness_function,
   crossover, mutation, Evolutionary_cycle, plot_bestfitness,
   plot_bestfitness_path, plot_pop_evolution,
   plot_pop_evolution2
4
5 function [time,X_best,fitness,f,k] = genetic(m)
6 tic
7
8 % Data Initialization
9 PP8_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,g,k);
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
37
38     % Genetic Operators
39     Offsprings_population = crossover(l_c, N_pop, fitness,
40         Population);
41     Offsprings_population = mutation(Offsprings_population, size
42         (Offsprings_population, 1), l_c);
43
44     % Fitness evaluation of the individuals in the sons'
45     population
46     [fitness_offsprings, ~, Offsprings_population] =
47         fitness_function(Offsprings_population, size(
48             Offsprings_population, 1), lb, ub, l_c, f, g, k+1);
49
50     [fitness, ~, Population] = fitness_function(Population, N_pop,
51         lb, ub, l_c, f, g, k);
52
53     % Survivors selection
54     [Population_new, fitness_new] = Evolutionary_cycle(
55         Population, N_pop, Offsprings_population, fitness,
56         fitness_offsprings);
57     Population = Population_new;
58     fitness = fitness_new;
59
60     % Decoding
61     X_pop = decoding(Population, N_pop, lb, ub, l_c);
62
63     if m == 1
64         % Plot - Current Generation
65         plot_bestfitness_path(X_pop, X_pop_old, pop_final)
66     end
67
68 end
69
70 % Final Population Fittest Individual
71 X_best = X_pop(1, :);
72
73 time = toc;
74
75 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);
75
76 % Plot - Generations Evolution - Standard Deviation and
    Mean
77 std_fitness(k+1) = std(fitness);
78 mean_fitness(k+1) = mean(fitness);
79 best_fitness(k+1) = fitness(1);
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', ...
86         'Approximate Solution: [%.5f , %.5f]\n', ...
87         'Objective function value: %.4f \n\n\n'], ...
88         k,X_best(1),X_best(2),fobj)
89 end
```

population_generation.m

```
1 %% population_generation.m
2 % Initial population generation
3
4 function [Population]=population_generation(N_pop,l_c)
5
6 Population1 = randi([0 1], N_pop, l_c);
7
8 % Eliminate Clones
9 Population = unique(Population1,'rows');
10
11 while size(Population1,1) > size(Population,1)
12     difference = size(Population1,1) - size(Population,1);
13     Population_add = randi([0 1], difference, l_c);
14     Population1 = [Population; Population_add];
15     Population = unique(Population1,'rows');
16 end
```

Evolutionary_cycle.m

```
1 %% Evolutionary_cycle.m
2 % Survivors Selection
3
4 function [Population_new,fitness_new] = Evolutionary_cycle(
    Population, ...
5     N_pop,Offsprings_population,fitness,
        fitness_offsprings)
6
7 % Selection (Fitness)
8 Natural_selection = [Population; Offsprings_population];
9 Natural_selection_fitness = [fitness; fitness_offsprings];
10
11 [Ordered_fitness,I] = sort(Natural_selection_fitness,1,'descend
    ');
12 Population_new=Natural_selection(I(1:N_pop),:);
13 fitness_new = Ordered_fitness(1:N_pop);
```

fitness_fuction.m

```
1 %% fitness_function.m
2 % Fitness Function
3 % Calls: decoding
4
5 function [fitness,X_real_ordered,population_ordered] =
    fitness_function(population,N_pop,lb,ub,l_c,f,g,k)
6
7 % Parameters
8 C = 0.5;
9 alfa = 3;
10 beta = 2;
11
12 % Decoding
13 X_real = decoding(population,N_pop,lb,ub,l_c);
14
15 % Average constraints violation
16 u=max(0,g(X_real(:,1),X_real(:,2)));
17
18 % Fitness evaluation
19 Aval = f(X_real(:,1),X_real(:,2)) + ((C*k)^alfa)*sum(u.^beta,2)
    ;
20 fitness = 1./Aval;
21
```

```
22 % Sorted Population (descend)
23 [fitness,position_pop]=sort(fitness,'descend');
24 X_real_ordered = X_real(position_pop,:);
25 population_ordered = population(position_pop,:);
26
27 end
```

decoding.m

```
1 %% decoding.m
2 % Genotype decoding (bits) to fenotype (real) - binary coding
  % emulating
3 % real representation
4
5 function [X_real] = decoding(Population,N_pop,lb,ub,l_c)
6
7 X_bin=zeros(N_pop,2);
8 X_real=zeros(N_pop,2);
9
10 for i=1:N_pop
11     for j=1:l_c/2
12         %x1
13         X_bin(i,1) = X_bin(i,1)+Population(i,j)*2^(l_c/2-j);
14
15         %x2
16         X_bin(i,2) = X_bin(i,2)+Population(i,j+l_c/2)*2^(l_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^(l_c/2) - 1)*X_bin(i
        ,2);
26 end
```

selection.m

```
1 %% selection.m
```

```
2 % Selection Operator
3
4 function [parent1, parent2, roulette] = selection(fitness,N_pop
    ,population,i, roulette)
5
6 if i == 1
7     % Selection probability for each individual
8     selection_prob = fitness/sum(fitness);
9
10    % Roulette intervals
11    roulette=zeros(N_pop,1);
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
19 selection_value = rand();
20 parent_position1 = find(selection_value<roulette,1);
21 parent1 = population(parent_position1,:);
22
23 % Select second parent
24 selection_value = rand();
25 parent_position2 = find(selection_value<roulette,1);
26
27 while parent_position2==parent_position1
28     selection_value = rand();
29     parent_position2 = find(selection_value<roulette,1);
30 end
31
32 parent2 = population(parent_position2,:);
```

crossover.m

```
1 %% crossover.m
2 % Crossover operator
3 % Calls: selection
4
5 function [offspring_pop] = crossover(l_c,N_pop,fitness,
    population)
6
7 offspring_pop = zeros(N_pop,l_c);
8 son1 = zeros(1,l_c);
```



```
9 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,
15         population,i,roulette);
16
17     % Mask
18     mask = randi([0,1],1,l_c);
19
20     for j=1:l_c
21         if mask(j) == 0
22             son1(j) = parent1(j);
23             son2(j) = parent2(j);
24         else
25             son1(j) = parent2(j);
26             son2(j) = parent1(j);
27         end
28     end
29
30     offspring_pop(i*2-1,:) = son1; % Son 1
31     offspring_pop(i*2,:) = son2;   % Son 2
32 end
```

mutation.m

```
1 %% mutation.m
2 % Mutation operator
3
4 function [offspring_pop] = mutation(offspring_pop,N_pop,l_c)
5
6 for i=1:N_pop
7     for j=1:l_c
8         probab_mut = rand();
9
10        if probab_mut <= 0.01
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
```

18 `end`

plot_bestfitness.m

```
1 %% File: plot_bestfitness.m
2 % Plot fittest individual for each generation
3
4 function [fig] = plot_bestfitness(ff,lb,ub)
5
6 fig = figure(1);
7 fcontour(ff,[lb(1) ub(1) lb(2) ub(2)])
8 title('$f(x_1,x_2)=2(x_1-2)^2+4(x_2-1)^2$', 'Interpreter','latex'
9      ')
10 xlabel('$x_1$',Interpreter='latex')
11 ylabel('$x_2$',Interpreter='latex')
12 dim = [0.15, 0.8, 0.1, 0.1];
13 const_str = {'Constraints:', '$2x_1+8x_2 \leq 6$', ' ', ...
14             '$2x_1 \geq 2x_2$',};
15 annotation('textbox',dim,'String', const_str,Interpreter='latex'
16            ',BackgroundColor='w')
```

plot_bestfitness_path.m

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

plot_pop_evolution.m

```
1 %% plot_pop_evolution.m
2 % Generation evolution plot
3
4 function [fig] = plot_pop_evolution(pop_evolution,X_pop,k,ff,lb
    ,ub)
5
6 fig = figure(2);
7 tiledlayout('flow')
8
9 a=round((k+1)/5);
10 for i=1:a:(1+4*a)
11     nexttile
12     fcontour(ff,[lb(1) ub(1) lb(2) ub(2)])
13     xlabel('$x_1$', 'Interpreter', 'latex')
14     ylabel('$x_2$', 'Interpreter', 'latex')
15     hold on
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')
25 hold on
26
27 scatter(X_pop(:,1),X_pop(:,2),"red")
28 title(['Generation ',num2str(k)], 'Interpreter', 'latex')
```

plot_pop_evolution2.m

```
1 %% plot_pop_evolution2.m
2 % Generation Evolution plot - standard deviation and mean
3
4 function [fig] = plot_pop_evolution2(std_fitness,mean_fitness,
    best_fitness,k)
5
6 fig = figure(3);
7
8 plot(0:k,std_fitness,"blue",'LineWidth',2) % Plot standard
    deviation
9 hold on
```

```
10 |
11 | plot(0:k,mean_fitness,"red",'LineWidth',2) % Plot mean
12 | hold on
13 |
14 | plot(0:k,best_fitness,"green",'LineWidth',2) % Plot best
    |     fitness
15 |
16 | title('Fitness','Interpreter','latex')
17 | xlabel('Generation','Interpreter','latex')
18 | xlim([0 k])
19 |
20 | legend('Standard Deviation','Mean','Best Fitness','Interpreter '
    |     , ...
21 |         'latex','Location', 'northeast')
```

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 point of minima obtained using a classical optimisation method, which is $f(5/3, 1/3) = 2$.

First Initial Population:
Number of generations: 40
Approximate Solution: [1.66950 , 0.33263]
Objective function value: 2.0000

Average results of 10 different initial populations:
Number of generations: 111
Approximate Solution: [1.66542 , 0.33365]
Objective function value: 2.0000
Maximum fitness value: 0.4996
Precision: 99.99972%
Average time: 0.8194s



