Optimizing Truss Structures Using Genetic Algorithms

Clearing Workspace

```
clc;
clear;
close all;
```

Initialization of global variables

```
global_variables
% global population_size VarMin VarMax
% global number_of_nodes nodes nodal_force fixed_nodes connectivity_matrix lengths_of_bars bars
% global E number_of_bars KG F U
```

Initialization Of variables bounds and Velocity Model (Young Model)

Loading Data

```
1 1 0 1 0
nodal_force = load('nodal_force')
nodal force = 6 \times 2
                    0
         0
      -2000
                    0
               -2000
         0
       2000
                    0
         0
                 2000
fixed_nodes = load('fixed_nodes')
fixed\_nodes = 1 \times 4
       2 11
                   12
number_of_nodes = size(nodes, 1)
number_of_nodes = 6
number_of_bars = sum(sum(connectivity_matrix))/2
number_of_bars = 13
bars = zeros(2, number_of_bars);
lengths_of_bars = zeros(1, number_of_bars);
```

Calculating the length of each bar

Creating Individuals data structure

every individual will have the structure (position, energy), where position is a number_of_bars-dimensional vector and energy if the deformation energy with respect to the lengths of bars.

```
agent.position=[];
agent.energy=[];
```

Initialization of genetic algorithm parameters

we will create variables:

- maximum_number_of_generations: The maximum number of generations
- population_size: The size of population (number of individuals in population)
- crossover_probability: Percentage of population that will be selected to crossover
- offspring_size: The number of individuals to crossover
- gamma: Extra range factor for crossover
- mutation_probability: Mutation probability
- number_of_mutantes: Number of mutants
- · mu: Mutation rate

```
maximum_number_of_generations=100

maximum_number_of_generations = 100

population_size=25

population_size = 25

crossover_probability=0.75

crossover_probability = 0.7500

offspring_size=2*round(crossover_probability*population_size/2)

offspring_size = 18

gamma=0.4

gamma = 0.4000

mutation_probability=0.1

mutation_probability = 0.1000

number_of_mutantes=round(mutation_probability*population_size)
```

```
number_of_mutantes = 3

mu=0.1

mu = 0.1000
```

now we will create the population with population_size is the number of agents (i.e individuals)

```
population=repmat(agent,population_size,1);
```

Selection method

a dialog will pop-up asking to select a selection method between:

- 1. Roulette Wheel
- 2. Tournament
- 3. Random

```
ANSWER=questdlg('Choose selection method:', 'Genetic Algorith', 'Roulette Wheel', 'Tournament' UseRouletteWheelSelection=strcmp(ANSWER, 'Roulette Wheel'); UseTournamentSelection=strcmp(ANSWER, 'Tournament'); UseRandomSelection=strcmp(ANSWER, 'Random');
```

if the user select "Roulette Wheel "we will add a variable beta to determinate the selection pressure.

if the user select " Tournament " we will add a variable TournamentSize to determinate the number of individuals putted into tournament.

```
if UseRouletteWheelSelection
    beta=8;
end

if UseTournamentSelection
    TournamentSize=3;
end
```

Initializing the population

- Generating positions randomly from the uniform distribution.
- Calculation the fitness for each generated position.

```
cost_function = @compliance;
for i=1:population_size
   population(i).position = unifrnd(VarMin, VarMax, [1 number_of_bars]);
   population(i).energy=cost_function(population(i).position);
end
```

```
costs = [population.energy];
[costs, SortOrder] = sort(costs);
population = population(SortOrder);
```

We will select the agent who has the best fitness (minimum fitness)

```
best_agent = population(1);
```

best cost value in each generation

```
best_agents=zeros(maximum_number_of_generations,1);
```

```
worst_cost = population(end).energy;
```

```
for it=1:maximum_number_of_generations
    % Calculate Selection Probabilities
    if UseRouletteWheelSelection
        P=exp(-beta*costs/worst_cost);
        P=P/sum(P);
    end
    % Crossover
    popc=repmat(agent,offspring_size/2,2);
    for k=1:offspring_size/2
        % Select Parents Indices
        if UseRouletteWheelSelection
            i1=RouletteWheelSelection(P);
            i2=RouletteWheelSelection(P);
        end
        if UseTournamentSelection
            i1=TournamentSelection(population, TournamentSize);
            i2=TournamentSelection(population, TournamentSize);
        end
        if UseRandomSelection
```

```
i1=randi([1 population_size]);
        i2=randi([1 population_size]);
    end
    % Select Parents
    p1=population(i1);
    p2=population(i2);
    % Apply Crossover
    [popc(k,1).position, popc(k,2).position]=Crossover(p1.position,p2.position,gamma,VarMi
    % Evaluate Offsprings
    popc(k,1).energy=cost_function(popc(k,1).position);
    popc(k,2).energy=cost_function(popc(k,2).position);
end
popc=popc(:);
% Mutation
popm=repmat(agent,number_of_mutantes,1);
for k=1:number_of_mutantes
    % Select Parent
    i=randi([1 population_size]);
    p=population(i);
    % Apply Mutation
    popm(k).position=Mutate(p.position,mu,VarMin,VarMax);
    % Evaluate Mutant
    popm(k).energy=cost_function(popm(k).position);
end
% Create Merged Population
population =[population
             popc
             popm]; % #ok
% Sort Population
costs=[population.energy];
[costs, SortOrder]=sort(costs);
population=population(SortOrder);
% Update Worst Cost
worst_cost=max(worst_cost,population(end).energy);
% Truncation
population=population(1:population_size);
costs=costs(1:population_size);
% Store Best Solution Ever Found
```

```
best_agent=population(1);

% Store Best Cost Ever Found
best_agents(it)=best_agent.energy;

% Show Iteration Information
%disp(['Iteration ' num2str(it) ': Best Cost = ' num2str(best_agents(it))]);
end
```

```
figure;
semilogy(best_agents, 'LineWidth', 2);
% plot(BestCost, 'LineWidth', 2);
xlabel('Generations');
title('Change in fitness over generations')
ylabel('Fitness');
grid on;
```

