## **ASSIGNMENT No: 08**

```
import random
from deap import base, creator, tools, algorithms
# Define the evaluation function (minimize a simple mathematical function)
def eval_func(individual):
   return (sum(x ** 2 for x in individual),) # Ensure tuple return
# DEAP setup
creator.create("FitnessMin", base.Fitness, weights=(-1.0,)) # Minimize function
creator.create("Individual", list, fitness=creator.FitnessMin)
toolbox = base.Toolbox()
# Define attributes and individuals
toolbox.register("attr_float", random.uniform, -5.0, 5.0) # Float values between -5 and 5
toolbox.register("individual", tools.initRepeat, creator.Individual, toolbox.attr_float, n=3) # 3D individual
toolbox.register("population", tools.initRepeat, list, toolbox.individual)
# Register evaluation function and genetic operators
toolbox.register("evaluate", eval_func)
toolbox.register("mate", tools.cxBlend, alpha=0.5)
toolbox.register("mutate", tools.mutGaussian, mu=0, sigma=1, indpb=0.2)
toolbox.register("select", tools.selTournament, tournsize=3)
# Create initial population
population = toolbox.population(n=50)
# Genetic Algorithm parameters
generations = 20
cxpb, mutpb = 0.5, 0.1 # Crossover and mutation probabilities
# Run the algorithm
for gen in range(generations):
   offspring = algorithms.varAnd(population, toolbox, cxpb, mutpb) # Apply genetic operators
   fits = list(map(toolbox.evaluate, offspring)) # Evaluate offspring
   for fit, ind in zip(fits, offspring):
       ind.fitness.values = fit # Assign fitness values
   population = toolbox.select(offspring, k=len(population)) # Select next generation
# Get the best individual after generations
best ind = tools.selBest(population, k=1)[0]
best_fitness = best_ind.fitness.values[0]
print("Best individual:", best_ind)
print("Best fitness:", best_fitness)
```

Best individual: [-0.0016319909959895042, -0.002012505361912262, -0.001641798406233576] Best fitness: 9.409074449427528e-06