Multi-objective optimisation implementation using Java

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Module 06-27818 and 27819: Advanced Aspects of Nature-Inspired Search and Optimisation (Ext)

- Existing EA/MOEA libraries and toolboxes
- 2 Introduction to MOEA Framework
- 3 How to define a new problem in MOEA
- 4 How to Evaluate algorithms?

Existing EA/MOEA frameworks and toolboxes

Matlab

- Matworks Global Optimisation Toolbox
- Genetic Algorithm Optimization Toolbox (GAOT)

Java

- ECJ: Java Evolutionary Computation Toolkit
- Watchmaker Framework for Evolutionary Computation
- JGAP: Java Genetic Algorithm Package
- MOEA framework ✓

• C/C++

- ParadisFO
- Evolving Objects (EO): an Evolutionary Computation Framework
- Open BEAGLE

Python

- PyGMO
- DEAP
- inspyred: Bio-inspired Algorithms in Python

- "A free and open source Java library for developing and experimenting with multiobjective evolutionary algorithms (MOEAs)" from MOEA framework web page
- Many state-of-the-art multiobjective evolutionary algorithms are included:
 - Genetic algorithms: NSGA-II and NSGA-III
 - Differential evolution
 - Particle swarm optimization
 - Genetic programming and Grammatical evolution
 - Click here for a list of all algorithms
- Other features:
 - Easy to use: minimum effor to define a problem, an algorithm and an experiment
 - Modular design: constructing new optimization algorithms from existing components
 - Very good documentation if you pay, but google "MOEA Framework Manual"

- Easier way:
 - Download the compiled binaries (Version 2.11) from here
 - Unzip and start Eclipse
 - Add all the *jar files into the libraries of your Java Build Path
- More comprehensive way (What we are going to use):
 - Download the source code (Version 2.11) from here
 - Unzip and start Eclipse
 - ullet Select File \longrightarrow Import
 - Select General → Existing Projects into Workspace
 - Select Set Root Directory → Click Browse button
 - Select your MOEAFramework-2.11 folder. Finally, click Finish

How to use MOEA?

General steps:

- Step 1: Define a new problem, extend the AbstractProblem class
- **Step 2**: Use Executor, Instrumenter or Analyzer classes, depends on your needs:
 - Executor: class for constructing and executing runs of an algorithm
 - Instrumenter: class for analyzing the performance of algorithms
 - Analyzer: class for analysing the resulting Pareto approximation set and how it compares against a known reference set.
- We shall see an example of solving a benchmark test function UF1 using MOEA later

- ZDT (Zitzler-Deb-Thiele) problems: 6 problems, e.g., ZDT1 -ZDT6
- DTLZ problems: 7 problems, e.g., DTLZ1 DTLZ7
- WFG problems: 9 problems, e.g., WFG1 WFG9
- CEC2009 competition problems: 10 problems, e.g., UF1 -UF10
- Other benchmarks problems: LZ, misc, etc.
- All problems are defined in MOEAFramework-2.11 \longrightarrow src \longrightarrow org.moeaframework.problem. + benchmark names
- MOEA Java Doc of the problems

• Example: CEC2009 competition UF1:

$$\min \begin{cases} f_1(\vec{x}) &= x_1 + \frac{2}{|J_1|} \sum_{j \in J_1} [x_j - \sin(6\pi x_1 + \frac{j\pi}{n})]^2 \\ f_2(\vec{x}) &= 1 - \sqrt{x_1} + \frac{2}{|J_2|} \sum_{j \in J_2} [x_j - \sin(6\pi x_1 + \frac{j\pi}{n})]^2 \end{cases}$$
(1)

where

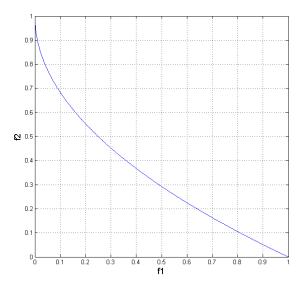
$$J_1 = \{j | j \text{ is odd and } 2 \le j \le n\}$$

$$J_2 = \{j | j \text{ is even and } 2 \le j \le n\}$$

$$0 \le x_1 \le 1, \ -1 \le x_i \le 1, \ i = 2, \cdots, n$$

MOEAs Test Functions: UF1

Pareto Front of UF1: $f_2 = 1 - \sqrt{f_1}$



We shall execute MOEA to solve UE1:

- Open MOEAFramework-2.11 \longrightarrow examples \longrightarrow (default package)
- Execute Example1.java
- Replace line 32 "NSGAII" with "Random"
- Execute it and compare the results from previous execution
- Explanation of the code

- Executor class: constructing and executing runs of an algorithm
- Required 3 pieces of information:
 - The problem (line 31)
 - The algorithm used to solve the problem (line 32)
 - the number of objective function evaluations allocated to solve the problem (line 33)
- Use run metohd to run this experiment and returns the resulting approximation set (line 35)
- Results, e.g., all Pareto optimal solutions produced by the algorithm during the run are stored in a NondominatedPopulation object (line 30)
- In NondominatedPopulation , there is an iterable class Solution , which stores the decision variables, objectives, constraints and attributes (line 39)

Code Explanation: Customising algorithms

- Executor executes algorithms with default parameters
- To set parameters of an algorithm: use withProperty (line 34)
- Each algorithm defines its own parameters
- See the API document for parameters of each algorithm

Defining a new problem in MOEA

- All problems in the MOEA Framework is defined by extending the AbstractProblem class
- Three things need to be done for defining a problem:
 - Characterizing a problem: number of decision variables and number of objectives
 - Defining the problems solution representation: Are they binary, integer, or real-number? what are their bounds?
 - Evaluation of the solutions: evaluate the solutions using objective functions
- Let's take a look at a new benchmark function DTLZ2

- Defined by Deb ,Thiele, Laumanns and Zitzler (DTLZ) in their paper
- ullet Can have m objectives and n (n>m) decision variables:

$$\min \begin{cases} f_1(\vec{x}) &= (1 + g(\vec{x}_m)) \prod_{i=1}^{m-1} \cos(x_i \frac{\pi}{2}) \\ f_2(\vec{x}) &= (1 + g(\vec{x}_m)) \sin(x_{m-1} \frac{\pi}{2}) \prod_{i=1}^{m-2} \cos(x_i \frac{\pi}{2}) \\ \dots \\ f_m(\vec{x}) &= (1 + g(\vec{x}_m)) \sin(x_1 \frac{\pi}{2}) \end{cases}$$
(2)

where \vec{x}_m is a vector of the remaining decision variables x_i , $i = \{m, \dots, n\}$, and

$$g(\vec{x}_m) = \sum_{r \in \vec{z}} (x_i - 0.5)^2$$
 (3)

and

$$-1 < x_i < 1, i = 1, \dots, n$$

MOEAs Benchamark Test Function DTLZ2

• For DTLZ2 with m=2 objectives and n=3 (n>m) decision variables:

$$\min \begin{cases} f_1(\vec{x}) = (1 + g(\vec{x}_m))\cos(x_1\frac{\pi}{2}) \\ f_2(\vec{x}) = (1 + g(\vec{x}_m))\sin(x_1\frac{\pi}{2}) \end{cases}$$
(4)

where \vec{x}_m is a vector of the remaining decision variables x_2 and x_3 , and:

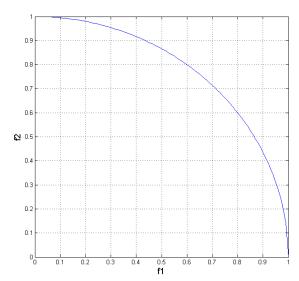
$$g(\vec{x}_m) = (x_2 - 0.5)^2 + (x_3 - 0.5)^2$$
 (5)

and

$$-1 \le x_i \le 1, \ i = 1, \cdots, n$$

MOEAs Test Functions: DTLZ2

Pareto Front of DTLZ2: $f_2 = \sqrt{1-f_1^2}$



Code example 3: Defining and solveing DTLZ2 problem (5 mins)

We will execute an example to solve the multi-objective benchmark problem DTLZ2:

- Open MOEAFramework-2.11 \longrightarrow examples \longrightarrow (default package)
- Execute Example4.java
- The code lines 70-73 are confusing:
 - Comment out line 70
 - Replace " numberOfVariables k " with " numberOfObjectives - 1 "
- Explanation of the code

- All problems in the MOEA Framework is defined by extending the AbstractProblem class (Line 34)
- Three things need to be done for defining a problem:
 - Characterizing a problem: use the constructor function to specify the number of decision variables and number of objectives (Lines 40-42)
 - Defining the problems solution representation: override newSolution methods in Soluton class to define:
 - Use Solution constructor to construct new solutions with the number of decision variables and number of objectives (lines 50-51)
 - Use a for-loop to iterate all variables to define variable types (binary, integer, real-number, etc.) and their bounds (lines 53-55)
 - Return the solution (line 57)
 - Evaluation of the solutions

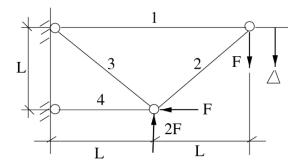
Code Explanation: How to define a new MOO problem

- All problems in the MOEA Framework is defined by extending the AbstractProblem class (Line 34)
- Three things need to be done for defining a problem:
 - Characterizing a problem
 - Defining problems solution representation
 - Evaluation of the solutions: Implement the evaluate method
 - Use getReal method in the EncodingUtils class to extract the decision variables from a solution (line 67)
 - Instantiate a double array to store the objective function values (line 68)
 - Use those those decision variables to evaluate the DTLZ2 objective functions, e.g., equations (2) - (3) (lines 72-87)
 - Finally we assign the objective function values to the solution (Line 89)

Exercise: multi-objective optimization in structural design

Problem: design of a four bar truss with two objectives:

- Minimise the volume of the truss
- Minimise its joint displacement



Exercise: multi-objective optimization in structural design

Mathematical formulation:

$$\min \begin{cases} f_1(\vec{x}) = L(2x_1 + \sqrt{2}x_2 + \sqrt{x_3} + x_4) \\ f_2(\vec{x}) = \frac{FL}{E}(\frac{2}{x_1} + \frac{2\sqrt{2}}{x_2} - \frac{2\sqrt{2}}{x_3} + \frac{2}{x_4}) \end{cases}$$

where

$$(F/\sigma) \le x_1 \le 3(F/\sigma)$$

$$\sqrt{2}(F/\sigma) \le x_2 \le 3(F/\sigma)$$

$$\sqrt{2}(F/\sigma) \le x_3 \le 3(F/\sigma)$$

$$(F/\sigma) \le x_4 \le 3(F/\sigma)$$
(6)

where x_i , $i=\{1,2,3,4\}$ are the cross sectional areas of the four bars, the acting force $F=10 \, \mathrm{kN}$, Young's modulus of elasticity $E=2 \times 10^5 \, \, \mathrm{kN/cm^2}$, the length $L=200 \, \mathrm{cm}$, the only nonzero stress component $\sigma=10 \, \mathrm{kN/cm^2}$.

- In your Eclipse MOEAFramework-2.11 project \longrightarrow examples \longrightarrow (default package): right click \longrightarrow New \longrightarrow Class \longrightarrow Type in your class name, e.g., ForBarProblemSolver
- After line 2 public class ForBarProblemSolver { ,
 add:
 public static class FourBar extends AbstractProblem {
 }
- Move your cursor to those place where Eclipse shows error with read underscore, e.g., FourBar, and auto-correct by selecting:
 - Add Constructor 'FourBar(int, int)'
 - Add Unimplemented Method
- Now you have a template to define the problem. Follow Example 4 to complete the exercise

Exercise 2: visualising Pareto front in Java (10 mins)

- Download my source code and open NSGAllexample.java
- I will explain how to use plotParetoFront class
- Plot the Pareto front of your four bar truss design problem

Multiobjective optimisation performance assessment

- We need to design performance metrics or indicators to evaluate the algorithms' performance.
- What characteristics should good Pareto optimal solutions have:
 - Minimize the distance of the Pareto front produced by the algorithm with respect to the true Pareto front if it is known
 - Maximize the spread of solutions found, so that we can have a as smooth and uniform Pareto front as possible.
 - Maximize the number of elements of the Pareto set (non-dominated solutions) found.
- A myriad of ways of metrics, but can be classified as three kinds for measuring:
 - Convergence: converge to true Pareto front
 - Spread: maintain diversity of solution mainly in objective space
 - Both Convergence and Spread

Code example 4: Performance assessment in MOEA

We shall execute an example to evaluate the performance of your algorithm using Analyzer:

- Open MOEAFramework-2.11 \longrightarrow examples \longrightarrow (default package) \longrightarrow Example2.java
- Add ", ''NSGAIII" " in line 31, just after " NSGAII "
- Execute Example2.java