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Parallel Multi-objective Optimization Method for Finding Complete Set of Weakly Efficient Solutions

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Problem statement

$$\min\{f(y) : y \in D\}, D = \{y \in \mathbb{R}^n : a_i \leq y_i \leq b_i, 1 \leq i \leq n\},$$

where $f(y)$ is a vector-function.

Solution of the problem is a set of non-dominated points (Slater set):

$$S(D) = \{y \in D : \nexists z \in D, f_i(z) < f_i(y), 1 \leq i \leq m\}$$

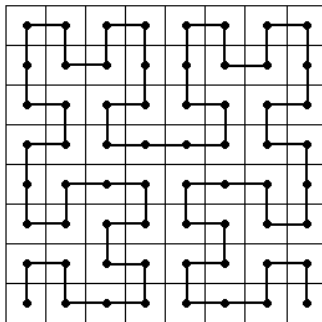
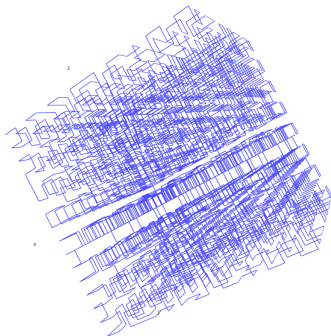
Assume objectives to satisfy Lipschitz condition in D :

$$|f_i(y_1) - f_i(y_2)| \leq L_i \|y_1 - y_2\|, y_1, y_2 \in D, 0 < L_i < \infty, 1 \leq i \leq m$$

Dimension reduction

Peano-type curve $y(x)$ allows to reduce dimension of the original multi-objective problem:

$$\{y \in \mathbb{R}^N : -2^{-1} \leq y_i \leq 2^{-1}, 1 \leq i \leq N\} = \{y(x) : 0 \leq x \leq 1\}$$
$$\min\{f(y) : y \in D\} = \min\{f(y(x)) : x \in [0, 1]\}$$



Scalarization technique

Multi-objective problem can be reduced to a scalar optimization problem using the following function:

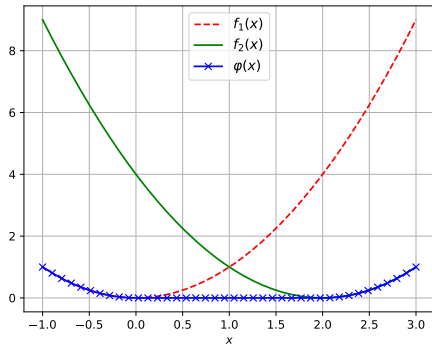
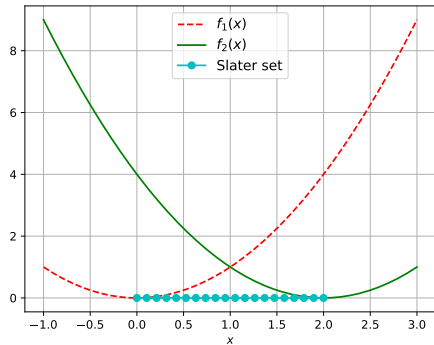
$$\begin{aligned} h(x, y) &= \min\{f_i(x) - f_i(y) : 1 \leq i \leq m\}, \\ \varphi(x) &= \max\{h(x, y) : y \in [0, 1]\}, x \in [0, 1]. \end{aligned}$$

If x^* is weakly effective point, then $\varphi(x^*) = 0$, in the opposite case $\varphi(x^*) > 0$, so the equivalent scalar problem is:

$$\varphi^* = \min\{\varphi(x) : x \in [0, 1]\}.$$

Scalarization technique

Example:



Optimization method

Optimization method generates search sequence $\{x_k\}$ and consists of the following steps:

- Step 1. Sort the search information (one-dimensional points) in increasing order.
- Step 2. Compute an approximation of the function $\varphi(x)$.
- Step 3. For each interval (x_{i-1}, x_i) compute quantity $R(i)$, called characteristic.
- Step 4. Choose an interval (x_{t-1}, x_t) with the greatest characteristic and compute objective $f(y(x))$ in the point chosen using the decision rule d :

$$x^{k+1} = d(t) \in (x_{t-1}, x_t)$$

- Step 5. If $x_t - x_{t-1} < \varepsilon$ stop the method.

Detailed description: Strongin R.G., Sergeyev Ya.D.: Global optimization with non-convex constraints. Sequential and parallel algorithms (2000), Chapter 7

Improved parallel optimization method

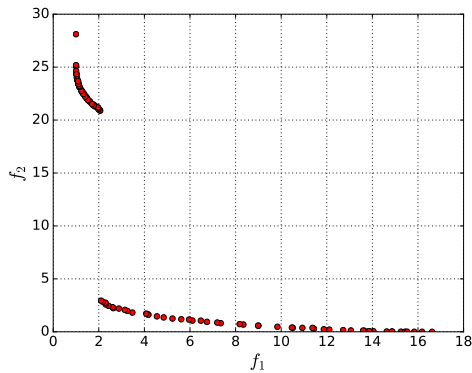
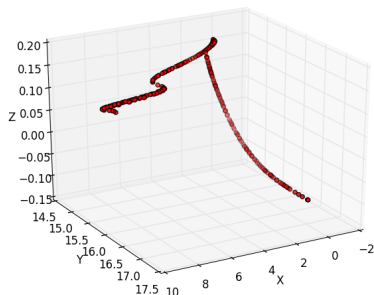
- ▶ *Local refinement.* Every q iterations ignore the characteristics $R(i)$ and perform calculation of the objective near the current running-minimum point of the $\varphi(x)$ approximation.
- ▶ *Parallelization by characteristics.* At the Step 4 choose p best intervals and generate p new points using rule $d(i)$. Then compute the objective $f(y(x))$ at these points in parallel exploiting p computation units.

In the best case (method doesn't generate redundant points compared to sequential one and wastes time only on computation of the objective) this scheme can give speedup in p times.

Results

The method was implemented on C++ language using OpenMP framework. All the experiments were carried out on 2xIntel Sandy Bridge E5-2660 2.2 GHz (total 16 cores) with 64Gb RAM.

Examples of numerical solutions:



Results

Number of iterations decreases when increasing number of threads p .

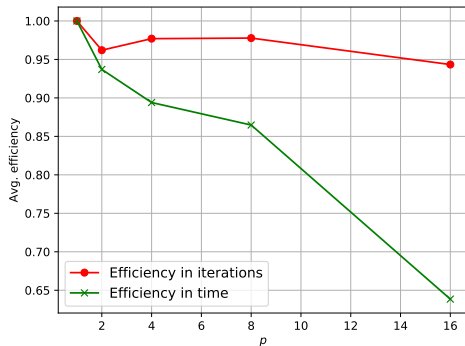
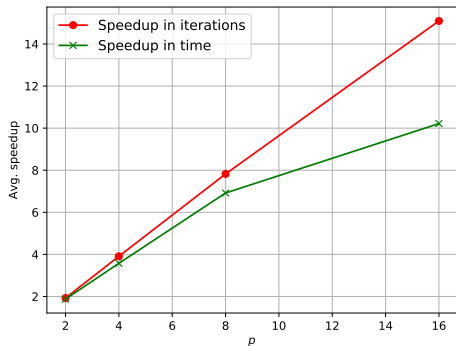
Problem	p				
	1	2	4	8	16
Markin-Strongin	1041(198)	516(198)	256(185)	131(197)	68(191)
Fonseca and Fleming 2d	1181(93)	636(99)	386(111)	176(95)	106(97)
Fonseca and Fleming 3d	5346(160)	3551(183)	1186(143)	606(153)	351(142)
Viennet problem	4896(276)	2156(273)	1226(270)	631(287)	286(274)
Poloni's function	3351(102)	1706(90)	856(88)	426(96)	201(99)

Results (speedup in time)

Each calculation of the objective takes about 8ms.

Problem	p				
	1(time, s)	2	4	8	16
Markin-Strongin	104.47	1.97	3.65	6.79	9.90
Fonseca and Fleming 2d	118.95	1.85	2.81	5.79	6.40
Fonseca and Fleming 3d	554.45	1.51	4.14	8.05	10.69
Viennet problem	1488.6	2.22	3.64	6.98	13.49
Poloni's problem	336.74	1.82	3.64	6.98	10.60

Results (speedup and efficiency)



Conclusion and future work

Already done:

- ▶ local refinement technique was introduced to the multi-objective method scheme;
- ▶ parallelism by characteristics was applied to the multi-objective method, efficiency metrics were collected.

Future work:

- ▶ compute quality metrics for the obtained solutions of the test problems;
- ▶ compare the method with other approaches;
- ▶ expand method to constrained problems;
- ▶ use MPI for distributed computation of objectives.



Q&A

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