

Lab 10 - Implementing and Understanding Non-dominated Sorting Genetic Algorithm (NSGA II)

CSE, SUSTech

Outline of This Lab

- Teaching Assistant
- What is Multiobjective Optimization (MOO) Problem?
- What is Non-dominated Solution?
- Non-dominated Sorting Genetic Algorithm (NSGA II)
- Test Problems
- Experimental Setup
- Illustrate the Results!

Teaching assistant

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What is Multiobjective Optimization (MOO) Problem?

- More than one objective to be optimised,
- with or without constraints.

$$\begin{aligned} \text{min/max } F(x) &= (f_1(x), f_2(x), \dots, f_m(x)) \\ \text{s.t. } g_j(x) &\geq 0, j = 1, 2, \dots, J \\ h_k(x) &= 0, k = 1, 2, \dots, K \\ x^{(L)}_i \leq x_i &\leq x^{(U)}_i, i = 1, 2, \dots, I \end{aligned}$$

where

- x is a vector of continuous, discrete or mixed variables.
- “s.t.” stands for “subject to”.
- m is the number of objectives.
- $x^{(L)}_i$ and $x^{(U)}_i$ refer to the lower bound and upper bound of x_i , respectively.

What is Non-dominated Solution?

-*Pareto (帕雷托) Dominance*

- x_a dominates x_b if
 - ✓ Solution x_a is no worse than x_b in all objectives.
 - ✓ Solution x_a is strictly better than x_b in at least one objective.
 - ✓ Denoted as $x_a \preceq x_b$ if minimisation.
- x_a dominates $x_b \Leftrightarrow x_b$ is dominated by x_a

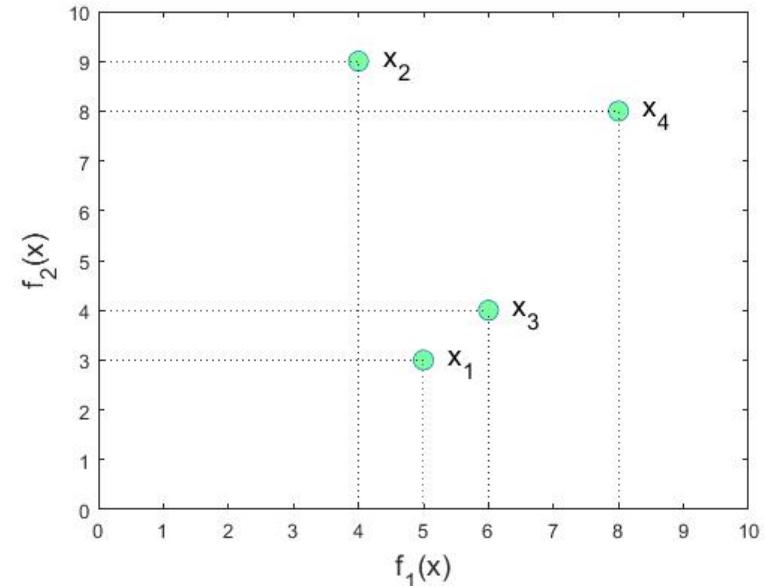


Figure 1: Example: minimise $F(x) = (f_1(x), f_2(x))$.

What is Non-dominated Solution?

-Pareto Front

- Among a set of solutions P , the non-dominated solution set is a set of solutions that are not dominated by any member of P .
- The non-dominated set of the entire feasible decision space is called the Pareto-optimal set.
- The boundary defined by the set of all points mapped from the Pareto optimal set to objective space is called the Pareto optimal front.

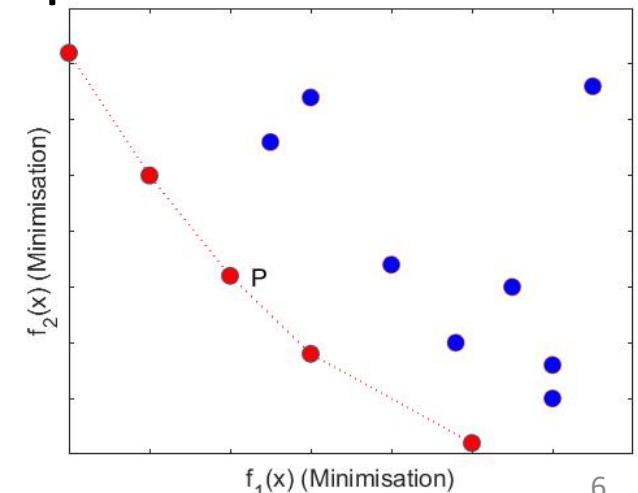


Figure 2: Pareto optimal: red points. Pareto optimal front: dashed red curve.

What is Non-dominated Solution?

-Pareto Optimal Solutions

- Pareto optimal set in the decision space (决策空间).
- Pareto optimal front in the objective space (目标空间).

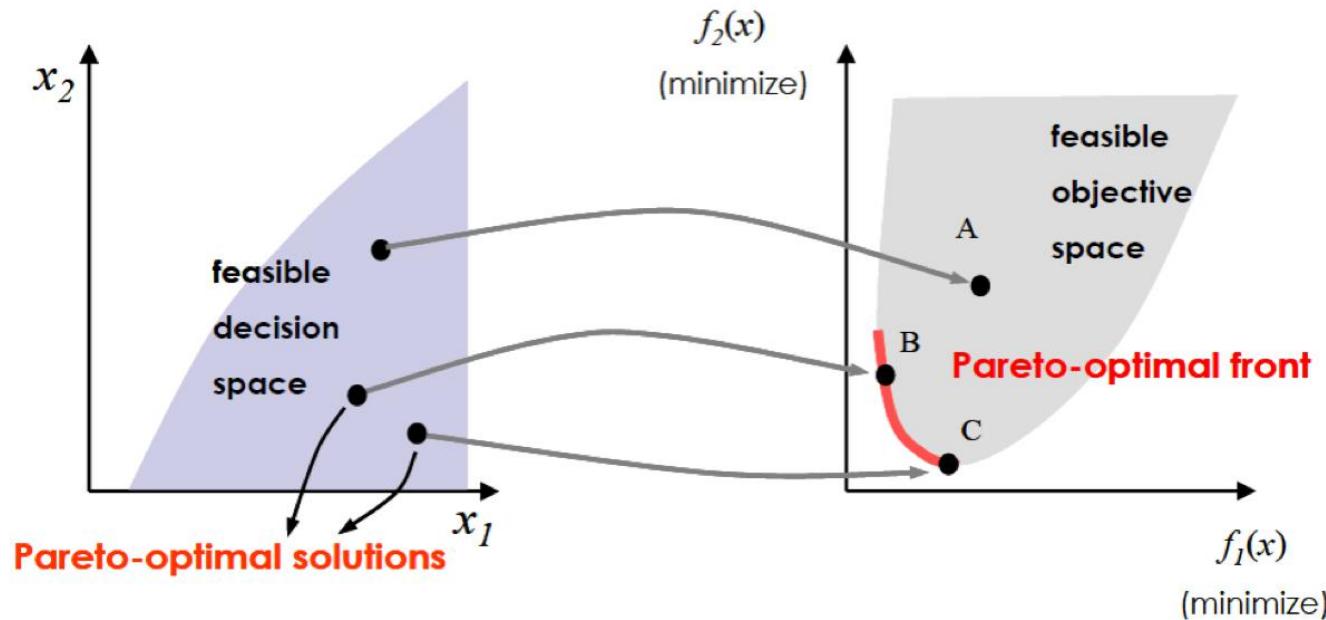


Figure 3: Image source: “Multi-Objective Optimization” by K. Deb.

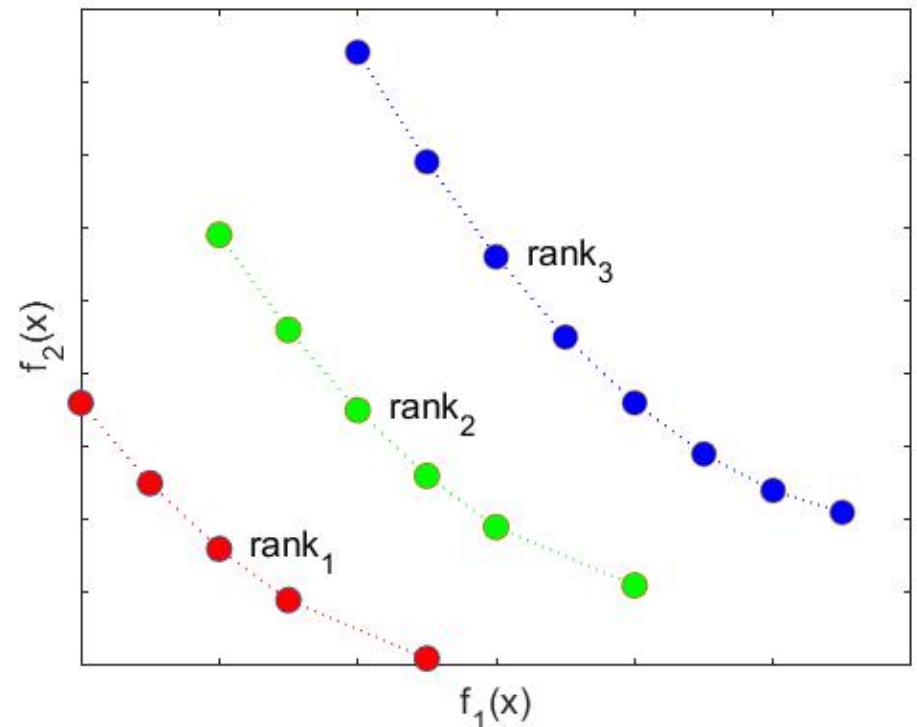
Non-dominated Sorting Genetic Algorithm (NSGA II)

-*Non-dominated Sorting*

Classify the solutions into a number of mutually exclusive non-dominated sets.

$$F = \bigcup_{i=1}^3 rank_i$$

[2] Kalyanmoy Deb et al. “A fast and elitist multiobjective genetic algorithm: NSGA-II”. In: IEEE transactions on evolutionary computation 6.2 (2002), pp. 182–197



Non-dominated Sorting Genetic Algorithm (NSGA II)

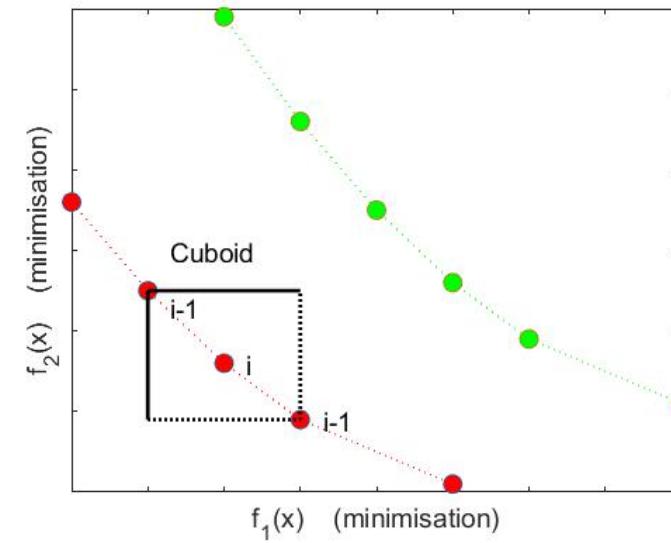
-Crowding Distance

- Determine crowding distance.
- Denotes half of the perimeter of the enclosing cuboid with the nearest neighbouring solutions in the same front.
- Estimation of the largest cuboid enclosing a particular solution (density estimation).

normalized m^{th} side distance:

$$d_i^m = \frac{d_{(i+1)(i-1)}^m}{f_{\max}^m - f_{\min}^m}$$

Figure 6: The crowding distance of the i^{th} solution in its front (red) is the average side-length of the cuboid (box) (or half of the perimeter of the enclosing cuboid with the nearest neighbouring solutions in the same front).



Non-dominated Sorting Genetic Algorithm (NSGA II)

-Comparing Solutions

- Crowding tournament selection
 - ✓ Assume that every solution has a non-domination rank and a local crowding distance.
 - ✓ A solution x_a wins a tournament against another solution x_b :
 - If the solution x_a has a better rank.
 - If they have the same rank but solution x_a has a larger crowding distance than solution x_b .

Non-dominated Sorting Genetic Algorithm (NSGA II)

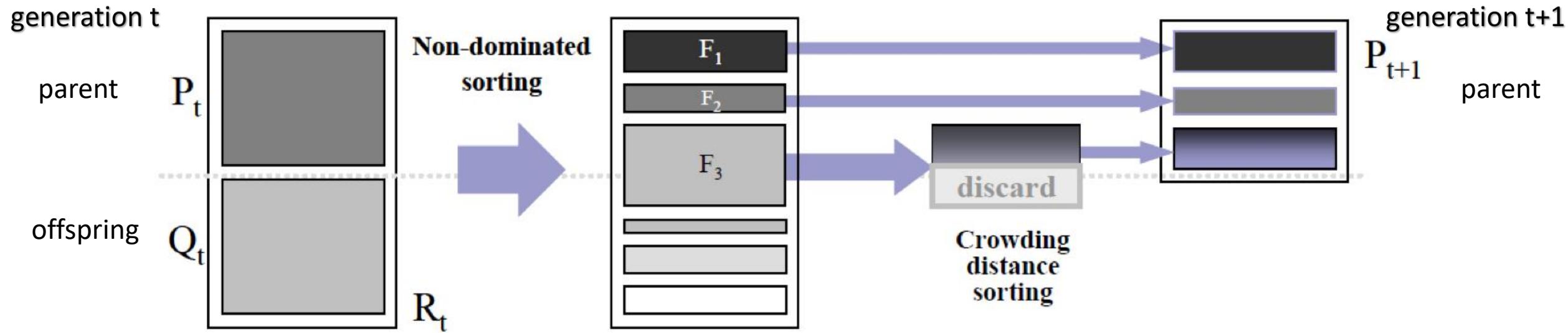


Figure 5: Image source: “Multi-Objective Optimization” by K. Deb.

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Elitist Non-Dominated Sorting GA (NSGA II)

-*main loop*

- Step 1
 - ✓ Create offspring population Q_t from P_t by using the crowded tournament selection, crossover and mutation operators
- Step 2
 - ✓ Combine parent P_t and offspring Q_t populations $R_t = P_t \cup Q_t$
 - ✓ Perform a non-dominated sorting to R_t and find different fronts F_i
- Step 3
 - ✓ Set new population $P_{t+1} = \emptyset$ and set $i = 1$
 - ✓ Until $|P_{t+1}| + |F_i| < N$, perform $P_{t+1} = P_{t+1} \cup F_i$ and increase i
- Step 4
 - ✓ Include the most widely spread solutions ($N - |P_{t+1}|$) of F_i in P_{t+1} using the crowding distance values

Test Problems

Problem	n	Variable bounds	Objective functions	Optimal solutions	Comments
ZDT1	30	[0, 1]	$f_1(\mathbf{x}) = x_1$ $f_2(\mathbf{x}) = g(\mathbf{x}) \left[1 - \sqrt{x_1/g(\mathbf{x})} \right]$ $g(\mathbf{x}) = 1 + 9 \left(\sum_{i=2}^n x_i \right) / (n - 1)$	$x_1 \in [0, 1]$ $x_i = 0,$ $i = 2, \dots, n$	convex
ZDT2	30	[0, 1]	$f_1(\mathbf{x}) = x_1$ $f_2(\mathbf{x}) = g(\mathbf{x}) \left[1 - (x_1/g(\mathbf{x}))^2 \right]$ $g(\mathbf{x}) = 1 + 9 \left(\sum_{i=2}^n x_i \right) / (n - 1)$	$x_1 \in [0, 1]$ $x_i = 0,$ $i = 2, \dots, n$	nonconvex
ZDT3	30	[0, 1]	$f_1(\mathbf{x}) = x_1$ $f_2(\mathbf{x}) = g(\mathbf{x}) \left[1 - \sqrt{x_1/g(\mathbf{x})} - \frac{x_1}{g(\mathbf{x})} \sin(10\pi x_1) \right]$ $g(\mathbf{x}) = 1 + 9 \left(\sum_{i=2}^n x_i \right) / (n - 1)$	$x_1 \in [0, 1]$ $x_i = 0,$ $i = 2, \dots, n$	convex, disconnected
ZDT4	10	$x_1 \in [0, 1]$ $x_i \in [-5, 5],$ $i = 2, \dots, n$	$f_1(\mathbf{x}) = x_1$ $f_2(\mathbf{x}) = g(\mathbf{x}) \left[1 - \sqrt{x_1/g(\mathbf{x})} \right]$ $g(\mathbf{x}) = 1 + 10(n - 1) + \sum_{i=2}^n [x_i^2 - 10 \cos(4\pi x_i)]$	$x_1 \in [0, 1]$ $x_i = 0,$ $i = 2, \dots, n$	nonconvex

Experimental Setup

Implement the Non-dominated Sorting Genetic Algorithm (NSGA II) and run it on the four test MOO problems with a maximum of 25 000 function evaluations for each run.

Illustrate The Results!

Plot 4 figures to display the found nondominated solutions for the four test problems, respectively .

- ✓ x-axis: f_1
- ✓ y-axis: f_2