

Main steps of the ABC algorithm

- (1) cycle = 1
- (2) Initialize the food source positions (solutions) $x_i, i = 1, \dots, SN$
- (3) Evaluate the nectar amount (fitness function fit_i) of food sources
- (4) **repeat**
- (5) Employed Bees' Phase
 - For each employed bee
 - Produce new food source positions v_i
 - Calculate the value fit_i
 - If new position better than previous position
 - Then memorizes the new position and forgets the old one.
 - End For.
- (6) Calculate the probability values p_i for the solution.
- (7) Onlooker Bees' Phase
 - For each onlooker bee
 - Chooses a food source depending on p_i
 - Produce new food source positions v_i
 - Calculate the value fit_i
 - If new position better than previous position
 - Then memorizes the new position and forgets the old one.
 - End For
- (8) Scout Bee Phase
 - If there is an employed bee becomes scout
 - Then replace it with a new random source positions
- (9) Memorize the best solution achieved so far
- (10) cycle = cycle + 1.
- (11) **until** cycle = Maximum Cycle Number

Algorithm 1: Pseudocode for ABC algorithm.

Main steps of the MOABC algorithm

- (1) cycle = 1
- (2) Initialize the food source positions (solutions) $x_i, i = 1, \dots, SN$
- (3) Evaluate the nectar amount (fitness fit_i) of food sources
- (4) The initialized solutions are sorted based on nondomination
- (5) Store nondominated solutions in the external archive EA
- (6) **repeat**
- (7) Onlooker Bees' Phase
 - For each onlooker bee
 - Randomly chooses a solution from EA
 - Produce new solution v_i by using expression (4.1)
 - Calculate the value fit_i
 - Apply greedy selection mechanism in Algorithm 3 to decide which solution enters EA
 - EndFor
- (8) The solutions in the EA are sorted based on nondomination
- (9) Keep the nondomination solutions of them staying in the EA
- (10) If the number of nondominated solutions exceeds the allocated size of EA
 - Use crowding distance to remove the crowded members
- (11) cycle = cycle + 1.
- (12) **until** cycle = Maximum Cycle Number

Algorithm 2: Pseudocode for MOABC algorithm.

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Greedy selection mechanism
If  $v_i$  dominates  $x_i$ 
    Put  $v_i$  into EA
Else if  $x_i$  dominates  $v_i$ 
    Do nothing
Else if  $x_i$  and  $v_i$  are not dominated by each other
    Put  $v_i$  into EA
    Produce a random number  $r$  drawn from a uniform distribution on the unit interval
    If  $r < 0.5$ 
        The the original solution is replaced by the new solution as new food source position.
        That means  $x_i$  is replaced by  $v_i$ .
    Else
        Do nothing
    End If
End If

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Algorithm 3: Greedy selection mechanism.

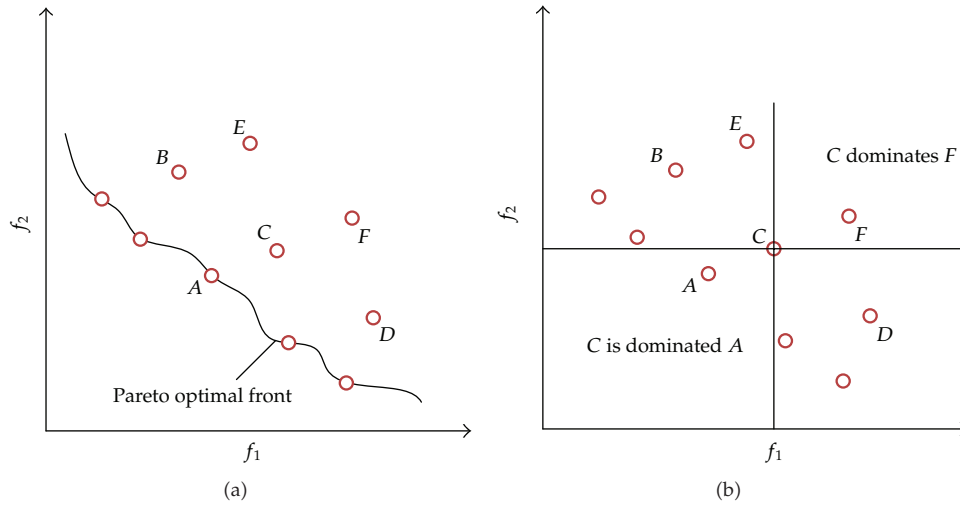


Figure 1: Illustrative example of Pareto optimality in objective space (a) and the possible relations of solutions in objective space (b).

Definition 2 (Pareto optimal). For (2.1), let $\mathbf{a} = (a_1, \dots, a_m)$, $\mathbf{b} = (b_1, \dots, b_m) \in R^m$ be two vectors, \mathbf{a} is said to dominate \mathbf{b} if $a_i \leq b_i$ for all $i = 1, \dots, m$, and $\mathbf{a} \neq \mathbf{b}$. A point $\mathbf{x}^* \in X$ is called (globally) Pareto optimal if there is no $\mathbf{x} \in X$ such that $\mathbf{F}(\mathbf{x})$ dominates $\mathbf{F}(\mathbf{x}^*)$. Pareto optimal solutions are also called efficient, nondominated, and noninferior solutions. The set of all the Pareto optimal solutions, denoted by PS, is called the Pareto set. The set of all the Pareto objectives vectors, $\text{PF} = \{\mathbf{F}(\mathbf{x}) \in R^m \mid \mathbf{x} \in \text{PS}\}$, is called the Pareto front [11, 12]. Illustrative example can be seen in Figure 1.