1 Algorithm Description

In this section, the algorithm is described. The population is being stored in an array of Ranks. Each Rank contains set of Individuals of a certain rank. This rank equals to the index of the corresponding Rank (or $non-domination\ level$) in the population array. In this algorithm Rank is implemented with a Dekart tree.

Algorithm 1 The function DETERMINERANK. It calculates rank of the new point p_n basing the ranks of points from $p \in P$ who dominate p_n

```
1: function DetermineRank(P, p_n)
2: R_n \leftarrow 0
3: for p \in P do
4: if x_p \leq x_{p_n} \wedge y_p \leq y_{p_n} then
5: R_n \leftarrow max(R_n, Rg(p) + 1)
6: end if
7: end for
8: return R_n
9: end function
```

Algorithm 2 The procedure ADDPOINT. On each step it splits tree of current rank into two parts: points, that should change rank (C_i) and points that should not. Then points, that have changed their rank on the previous steps, are being added to the remainder. The proof is given in Theorem 1

```
1: procedure ADDPOINT(P, p_n)
         if p_n \in P then return
 3:
 4:
              R_n \leftarrow \text{DetermineRank}(P, p_n)
 5:
 6:
              p_0 \leftarrow p_n
              C_{-1} \leftarrow \{p_n\}
 7:
              C_0 \leftarrow \{p : Rg(p) = R_n \land p_n \prec p\}
 8:
 9:
              while |C_i| \neq 0 do
10:
                   if \nexists P[R_n+i] then
                        P[R_n + i] \leftarrow C_i
11:
                       return
12:
                   end if
13:
                   P[R_n + i] \leftarrow \text{CutTree}(P[R_n + i], C_i)
14:
                   P[R_n + i] \leftarrow \text{AddTree}(P[R_n + i], C_{i-1})
15:
                   p_{i+1} \leftarrow (\min c \in C_i x_c, \min c \in C_i y_c)
16:
                   i \leftarrow i + 1
17:
                   C_i \leftarrow \{p : Rg(p) = R_n + i \land p_i \prec p\}
18:
              end while
19:
              P[R_n + i] \leftarrow \text{AddTree}(P[R_n + i], C_{i-1})
20:
         end if
21:
22: end procedure
```

2 Proof

Lemma 1. If:

$$C = \{c : Rg(c) = R\},\tag{1}$$

$$p_0 = (min_{c \in C}c_x; min_{c \in C}c_y), \tag{2}$$

$$\nexists p': Rg(p') = R \quad and \quad x_{p'} \in [min_{c \in C}c_x; max_{c \in C}c_x]. \tag{3}$$

Then:

$$D_{p_0} = \{d : p_0 \prec d \land Rg(d) > R\} = D_C = \bigcup_{c \in C} \{d : c \prec d\}$$
 (4)

Where D_e is a set of elements, dominated by e (or at least by one member of e, if this is a set).

Proof. 1. By definition of p_0 ,

$$p_0 \leq c, \forall c \in C$$

That leads, by the transitivity of \leq , to the following:

$$c \in C \prec d \Rightarrow p_0 \prec d$$

2. Let's assume the following:

$$p_0 \prec d, \tag{5}$$

(5) can lead to the following cases:

1. $x_d = x_{p_0}, y_d > y_{p_0};$

According to (2),

$$\exists c_1 \in C : x_{c_1} = x_{p_0}$$

That means, that $x_{c_1} = x_d$. According to (6), $y_{c_1} >= y_d$. That means, that

$$d \leq c_1 \Rightarrow Rg(c_1) >= Rg(d) >^{(4)} R$$

This contradicts with (1).

2. $x_d > x_{p_0}, y_d = y_{p_0};$

The proof is the same as for the previous case.

3. $x_d > x_{p_0}, y_d > y_{p_0};$

$$Rg(d) >^{(4)} R \Rightarrow \exists p \notin C : Rg(p) = R, p \prec d$$

Let's introduce the following variables:

$$c_1: c_1 \in C, x_{c_1} = x_{p_0}$$

$$c_2: c_2 \in C, y_{c_1} = y_{p_0}$$

According to (3),

$$x_p < x_0 \lor y_p < y_0$$

which means that $x_p < x_{c_1}$, and, according to the definition of rank, $y_p > y_{c_1}$. But $x_d > x_{p_0} = x_{c_1}$. Therefore:

$$p \prec d \Rightarrow c_1 \prec d$$

It contradicts with (6). The similar proof is applicable to c_2 .

Let's define Rg(p) as rank of p before point addition, and Rg'(p) as rank of p after point addition. Let's define R_i :

$$F_i: \{ \forall f \in F_i: Rg(f) = F_i, p_i \not\prec f \}$$

Theorem 1. Point n was added. $Rg'(n) = R_0$. The following statements are applicable for any iteration of point addition algorithm:

1. $\forall i > 0 : R_i = R_0 + i;$

- 2. $\exists p_i, C_i : \{ \forall c \in C_i : Rg(c) = R_i, p_i \prec c \};$
- 3. $Rg'(C_i) = R_i + 1$;
- 4. $Rg'(F_i) = R_i$;
- 5. $p_{i+1} = (min_{c \in C_i} c_x; min_{c \in C_i} c_y).$

Proof. The proof will be by induction.

- 1. Base.
 - (a) $p_0 = n$;
 - (b) $\exists C_0 : \{ \forall c \in C_0 : Rg(c) = R_0, n \prec c \};$
 - (c) $Rg'(C_0) = R_0 + 1 = R_i + 1$;
 - (d) $Rg'(F_0)$ won't be changed;
 - (e) $p_{i+1} = (min_{c \in C_0} c_x; min_{c \in C_0} c_y).$
- 2. Induction step.
 - (a) $p_{i+1} = (min_{c \in C_i} c_x; min_{c \in C_i} c_y);$
 - (b) $Rg'(C_i) = R_i \Rightarrow^{lemma1} \forall d : Rg(d) = R_i, p_i \prec d : Rg'(d) = Rg(C_i) + 1 = R_i + 1;$
 - (c) $Rg'(F_i)$ won't be changed.

3 Running time complexity

3.1 Running time of DetermineRank

This function iterates over the whole population and compares the new point against every one of the existing points. This gives us N comparsions. Each comparsion costs O(k) operations, because it's required to compare k components of each individual. Therefore, the running time complexity of DetermineRank is O(N*k).

3.2 Running time of AddPoint

In the worst case, the new point will be added to the first non-domination level. In this case we'll have to update all levels during the addition of this point. Let's introduce L as the average size of a non-domination level and M as the total number of non-domination levels. The *while* loop can give us up to M iterations. Obviously,

$$N \ge M * L \to M \le N/L \tag{7}$$

Inside each iteration:

- 1. Procedure CutTree costs $O(\log(L))$ (according to the properties of Dekart tree);
- 2. Procedure AddTree costs $O(\log(L))$ (according to the properties of Dekart tree);
- 3. Calculation of C_i costs O(L*k) (because it's required to check all the points of the next rank).

So, it's $M*(\log(L)+\log(L)+L*k)=2*M*\log(L)+M*L*k\leq 2*N*\log(L)/L+N*k$ in total. This function reaches its maximum at L=2. Therefore, the running time complexity of AddPoint will be O(N+N*k)=O(N*k).

3.3 Total running time

So, the total running time of the proposed algorithm is O(N*(k+1)+N*k) = O(N*k)