# Clusterability

### 1 Problem Statement

#### 1.1 Notation

For any set  $B \subset X$ , we denote c(B) as the center of B which is defined as the average of points in B. Radius of the set B is defined as  $r(B) = \max_{x \in B} |x - c(B)|$ .

**Definition 1** (Niceness assumption). Given a set  $\mathcal{X}$ , we say that a partition of  $\mathcal{X}$ ,  $C_1, ..., C_k$  is  $(\lambda, \nu)$ -nice if the following conditions hold. There exist sets  $B_1, ..., B_k \subset \mathcal{X}$  such that for every  $i \in [k]$ , there exists  $j_i \in [k]$  such that  $B_i \subset C_{j_i}$ .

- Separation: For all  $i, j \in [k], |c(B_i) c(B_j)| \ge \nu \cdot \max\{r(B_i), r(B_j)\}$
- Sparse Noise: For any ball  $B \subset \mathcal{X}$  for which  $r(B) \leq \lambda \cdot \max_{i \in [k]} r(B_i), |B \cap \{X \setminus \bigcup_{i \in [k]} B_i\}| \leq 1$

# 2 Algorithm

## 2.1 Based on the knowledge of $\min |B_i|$

The algorithm gets as input  $\mathcal{X}$  and the number of points in the smallest cluster min  $|B_i| = t(\text{say})$ . This algorithm works in two phases.