

# Parallel $k$ -Means in Theory in Practice

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# $k$ -Means

Input:

- Number of desired means  $k \in \mathbb{N}$
- Set of points  $P \subset \mathbb{R}^d$  (or multiset)

Output:

- Set of  $k$  means  $C = \{c_1, \dots, c_k\}$

# Lloyd's Heuristic

- Additional input: Seeding  $C$ 
  - 1: **repeat**
  - 2:     **for**  $x \in P$  **do**
  - 3:          $a[x] \leftarrow \arg \min_{c \in C} \text{dist}(x, c)$
  - 4:     **for**  $c \in C$  **do**
  - 5:          $c \leftarrow \text{mean}(\{x \in P \mid a[x] = c\})$
  - 6: **until**  $C$  did not change in last iteration

# Lloyd's Heuristic

- 1:  $C \leftarrow \{\text{random } p \in P\}$
- 2: **while**  $|C| < k$  **do**
- 3:      $C \leftarrow C \cup \{\text{random } p \in P, \text{ sampled } \sim \text{dist}(p, C)^2\}$

# Implementation in SQL (1/2)

$C \leftarrow \{\text{random } p \in P\}$ :

```
1  SELECT ARRAY[(
2      SELECT CAST($expr_point AS DOUBLE PRECISION[])
3      FROM $rel_source
4      WHERE $expr_id = (
5          SELECT weighted_sample($expr_id, 1)
6          FROM $rel_source
7      )
8  )]
```

# Implementation in SQL (2/2)

$C \leftarrow C \cup \{\text{random } p \in P, \text{ sampled } \sim \text{dist}(p, C)^2\}$

```
1 | SELECT centroids || $expr_point
2 | FROM $rel_source
3 | WHERE $expr_id = (
4 |     SELECT weighted_sample(
5 |         $expr_id, (
6 |         closest_column(
7 |             centroids,
8 |             $expr_point,
9 |             fn_squared_dist
10 |         )).distance
11 |     ) FROM $rel_source)
```

# k-means||

- 1:  $C \leftarrow \{\text{random } p \in P\}$
- 2:  $\Phi_0 \leftarrow \Phi(C)$
- 3: **for**  $O(\log \Phi_0)$  times **do**
- 4:      $C' \leftarrow \{p \in P, \text{ each with prob. } \frac{\ell \cdot \text{dist}(p, C)^2}{\Phi(C)}\}$
- 5:      $C \leftarrow C \cup C'$
- 6: **for**  $c \in C$  **do**
- 7:      $w_c \leftarrow \# \text{ points in } P \text{ “assigned” to } c$
- 8: Run (weighted)  $k$ -means++ on  $C$