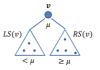
# **VP-tree in Search of Nearest** Neighbors within a Given Radius

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#### The Idea of Constructing a VP-Tree

- A node in VP-Tree contains:
- $-\mu = median(\{u \in S(v) | distance(u, v)\}),$ where S(v) is the subtree rooted in v} LS(v) - link to left subtree LS(v) embracing
- $\{u \in S(v) \setminus \{v\} | distance(u, v) < \mu\}$ link to right subtree RS(v) embracing  $\{u \in S(v) \backslash \{v\} | distance(u,v) \geq \mu\}$

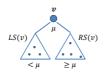


- Each point in D is stored only once in VP-Tree.
- Idea of how to select a point from D to be stored in the root of the VP-tree:
- A point in the root of the VP-tree, say point  $\nu$ , should be the one with the maximal spread of its distances to all points in D.
- Idea of how to select a point to be the root of a subtree covering a subset D' of points in D:
- A point in the root of this subtree, say point v, should be the one with the maximal spread of its distances to all points in D'.

#### **Practical Construction of a VP-Tree**

- A node in VP-Tree contains:

  - $-\mu = median(\{u \in S(v) | distance(u, v)\}),$
  - where S(v) is the subtree rooted in v}
     link to left subtree LS(v) embracing
  - find to left subtree LS(v) either acting  $\{u \in S(v) \setminus \{v\} | distance(u, v) < \mu\}$  link to right subtree RS(v) embracing  $\{u \in S(v) \setminus \{v\} | distance(u, v) \ge \mu\}$

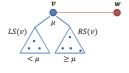


- Practical selection of a point from (a subset D' of) D to be stored in the root of a (sub-)tree:
- A random sample  $S_1$  of points from (subset D' of) D constitutes a set of candidates to be stored in the root of the (sub-)tree.
- Their spreads of distances are calculated with respect to another random sample  $S_2$  of points from (subset D' of) D.
- The candidate point  $\nu$  from  $S_1$  with the maximal spread of its distances to the points in the sample  $S_2$  is stored in the root of the (sub-tree).
- The real median of this point v is calculated based on its distances to all points in (subset D' of) D, and is also stored in the root of the (sub-)tree.

#### k/k\*-NN Search in VP-Tree

- A node in VP-Tree contains:

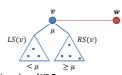
  - $\begin{array}{l} \cdot v \in D \\ -\mu = median(\{u \in S(v) | distance(u,v)\}), \\ \text{where } S(v) \text{ is the subtree rooted in } v\} \\ -\text{ link to left subtree } LS(v) \text{ embracing} \\ \{u \in S(v)\setminus \{v\}| distance(u,v) < \mu\} \\ -\text{ link to right subtree } RS(v) \text{ embracing} \\ \{u \in S(v)\setminus \{v\}| distance(u,v) \geq \mu\} \end{array}$



- Search for k/k+-NN of point u within  $\varepsilon$  radius in node v of VP-Tree
- Cond. 1:  $distance(w, v) \mu \ge \varepsilon$ . If true, then for each point u in LS(v),  $distance(w,v) - distance(u,v) > distance(w,v) - \mu \ge \varepsilon$ . Thus,  $distance(w,v) - distance(u,v) > \varepsilon$ , so LS(v) does not contain  $k/k^+$ -NN(w)
- Cond. 2:  $\mu distance(w, v) > \varepsilon$ . If true, then for each point u in RS(v),  $\begin{array}{l} \textit{distance}(u,v) - \textit{distance}(w,v) \geq \mu - \textit{distance}(w,v) > \varepsilon. \text{ Thus,} \\ \textit{distance}(u,v) - \textit{distance}(w,v) > \varepsilon, \text{so } RS(v) \text{ does not contain } k/k^* - NN(w) \end{array}$ within a radius

# Improved k/k+-NN Search in VP-Tree...

- A node in VP-Tree contains:
- $-\mu = median(\{u \in S(v) | distance(u, v)\}),$ where S(v) is the subtree rooted in v}
- link to left subtree LS(v) embracing  $\{u \in S(v) \setminus \{v\} | distance(u, v) < \mu\}$  link to right subtree RS(v) embracing  $\{u \in S(v) \setminus \{v\} | distance(u, v) \ge \mu\}$



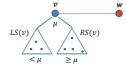
- Search for k/k\*-NN of point u within  $\varepsilon$  radius in node v of VP-Tree:
- distance(w,v), Cond. 1:  $distance(w,v) \mu \ge \varepsilon$ . If true, then for each point u in LS(v),  $\begin{array}{l} \textit{distance}(w,v) - \textit{distance}(u,v) > \varepsilon, \text{so kNN}(w) \text{ is not in } LS(v) \text{ within } \varepsilon \text{ radius.} \\ - \text{Cond. } 2: \mu - \textit{distance}(w,v) > \varepsilon. \text{ if true, then for each point } u \text{ in } RS(v), \\ \textit{distance}(u,v) - \textit{distance}(w,v) > \varepsilon, \text{so kNN}(w) \text{ is not in } RS(v) \text{ within } \varepsilon \text{ radius.} \\ \end{array}$
- Improved search for k/k+-NN of point u within  $\varepsilon$  radius in node v of VP-Tree - distance(w, v),
  - Cond. 1':  $distance(w, v) left\ bound > \varepsilon$ , where  $left\ bound\$ is the maximum of the distances from point v to all points in LS(v).

    - Cond. 2':  $right\_bound - distance(w, v) > \varepsilon$  , where  $right\_bound$  is the
  - minimum of the distances from point v to all points in RS(v)

### Improved k/k+-NN Search in VP-Tree

- A node in VP-Tree contains:
  - $v \in D$   $\mu = median(\{u \in S(v) | distance(u, v)\}),$

  - $\mu$  measum{u = \lambda v \neq \left(u \) = \lambda v \text{ measure}(u, \neq \left\), where S(v) is the subtree rooted in v} link to left subtree LS(v) embracing  $\{u \in S(v) \setminus \{v\} | distance(u, v) < \mu\}$  link to right subtree RS(v) embracing  $\{u \in S(v) \setminus \{v\} | distance(u, v) \ge \mu\}$



- Improved search for k/k+-NN of point u within  $\varepsilon$  radius in node v of VP-Tree: - distance(w, v),
- Cond. 1':  $distance(w, v) left\_bound > \varepsilon$ , where  $left\_bound$  is the maximum of the distances from point v to all points in LS(v). - Cond. 2':  $right\_bound - distance(w, v) > \varepsilon$ , where  $right\_bound$  is minimum of the distances from point v to all points in RS(v).
- **Example.** Let  $\varepsilon = 1$ ,  $\mu = 10.5$ , left bound(v) = 8.5, right bound(v) = 12 and  $\frac{distance(w,v)}{distance(w,v)} = 10. \ Then, \ \frac{distance(w,v)}{distance(w,v)} - \frac{1}{\epsilon} \frac{bound(v)}{s} \varepsilon \ and \ \frac{right\_bound(v)}{s} - \frac{1}{\epsilon} \frac{distance(w,v)}{s} - \frac{1}{\epsilon} \frac{distance(w,v)}{s} = \frac{1}{\epsilon} \frac{distance(w,v)}{s} - \frac{1}{\epsilon} \frac{distan$ neighbor of w within the  $\varepsilon$  radius.

# References

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