



K Dimensional Tree

Group 8:

Cát Văn Tài

Lý Hoàng Thuận

Nguyễn Hữu Hưng

Deep Image Retrieval

The goal of this project is deep image retrieval, that is learning an embedding (or mapping) from images to a compact latent space in which cosine similarity between two learned embeddings correspond to a ranking measure for image retrieval task.

Validate

Upload

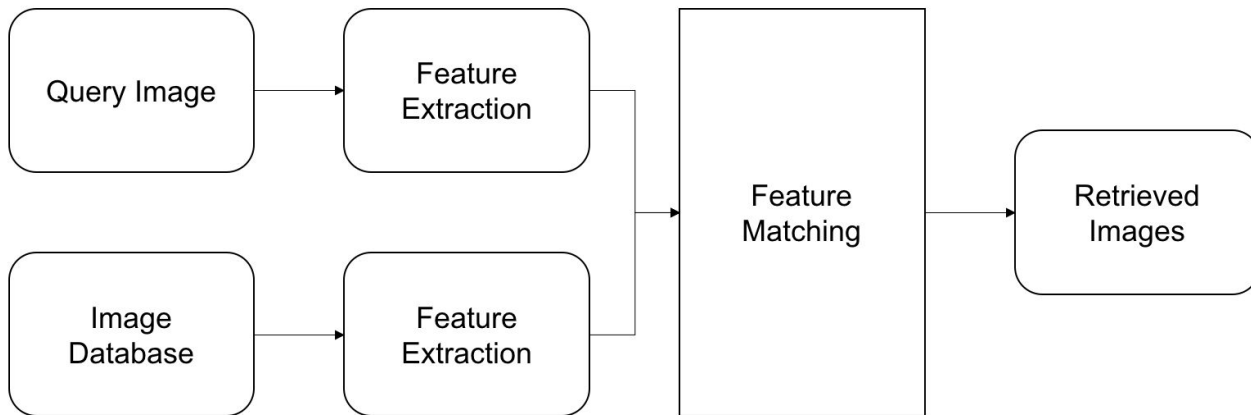
Pick an image to validate

Oxford

Paris



introduce IR

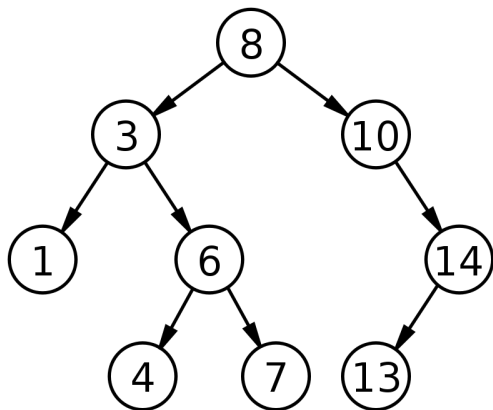


The pipeline for IR

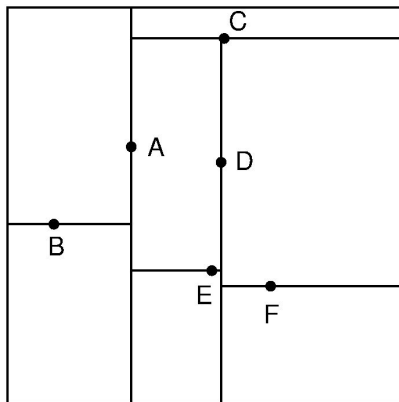
- Given 100.000 samples with n_{features} is 128. Then, input a any sample and convert it to vector 128. Let's find the nearest samples compare with the input.
- Solutions for feature matching: linear search, **k-d tree**, etc.

k-d tree

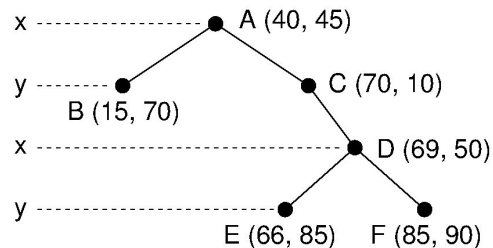
- A k-d tree is a binary search tree for partitioning data
- It is used for various applications like nearest points, range search



The representation of binary tree

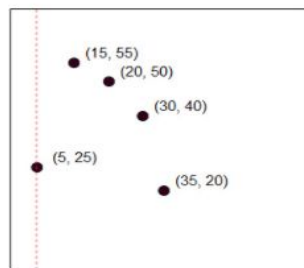


The representation of k-d tree

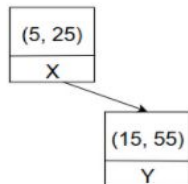
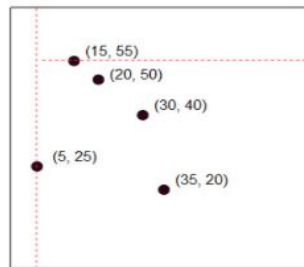


construction

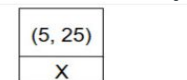
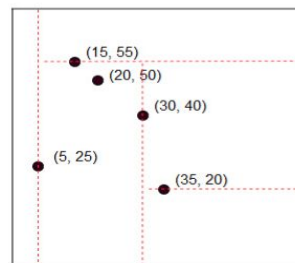
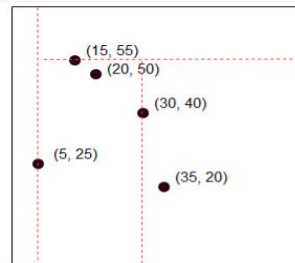
There are many possible ways to choose axis-aligned splitting planes, so there are many different ways to construct k-d trees.



Insert
(5, 25)



Insert
(15, 55)



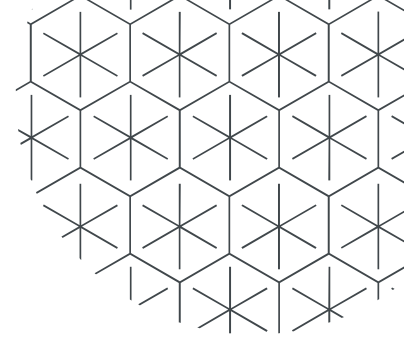
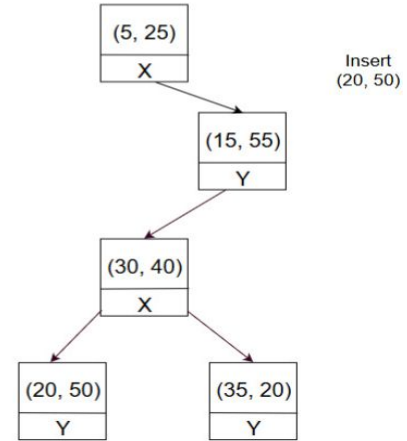
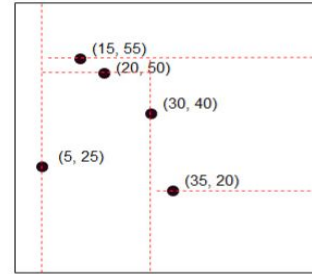
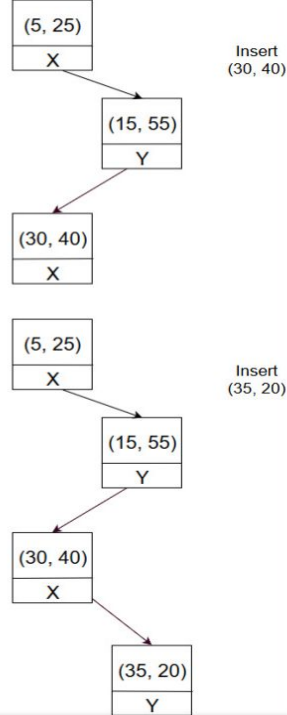
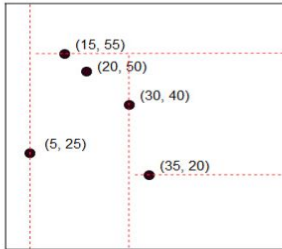
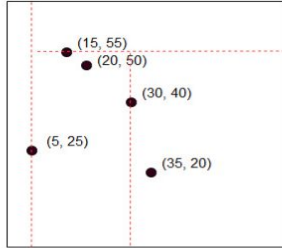
Insert
(30, 40)



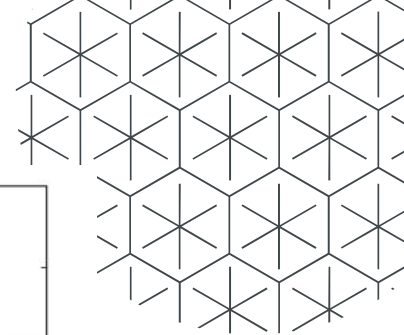
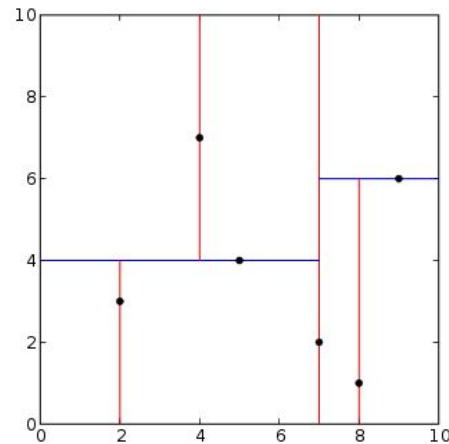
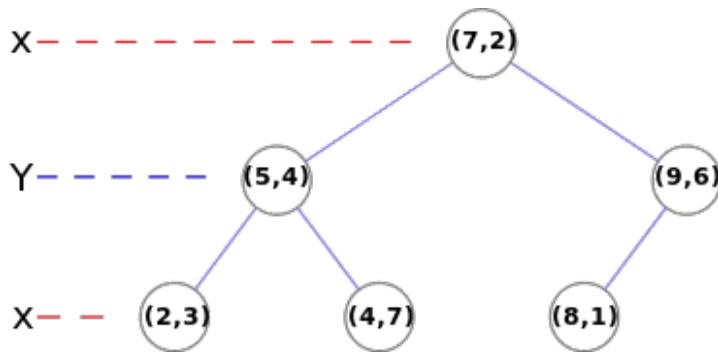
Insert
(35, 20)



construction



construction

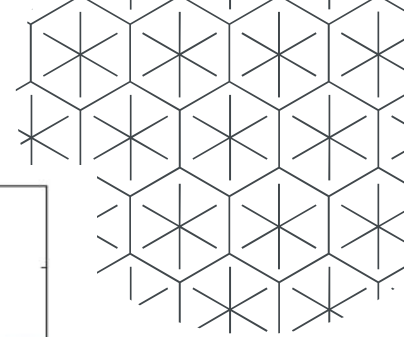
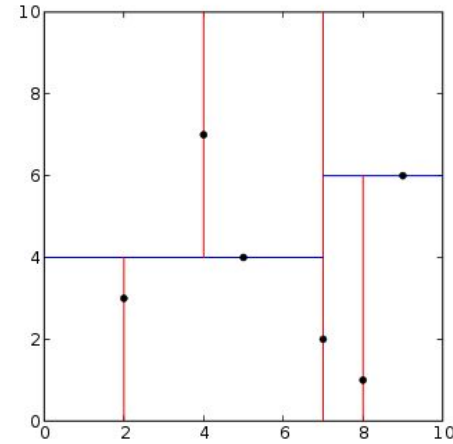
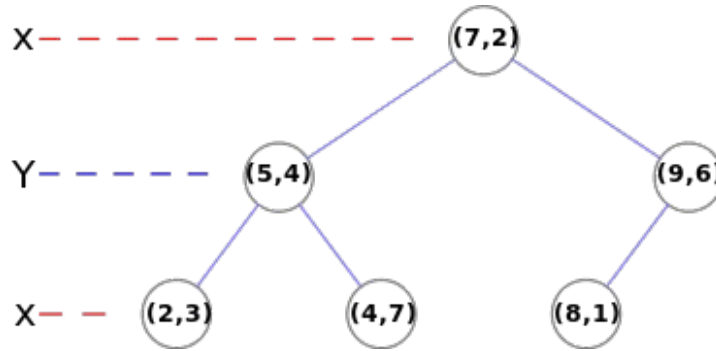


The **canonical method** of k-d tree construction has the following constraints:

- As one moves down the tree, one cycles through the axes used to select the splitting planes
- Points are inserted by selecting the median of the points being put into the subtree

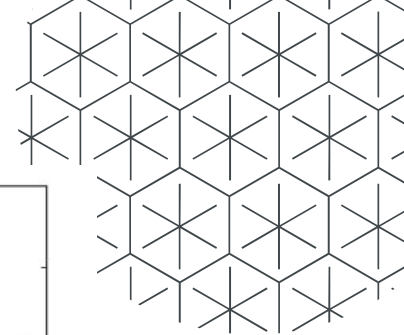
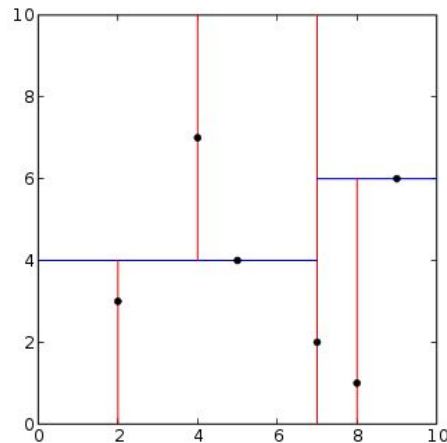
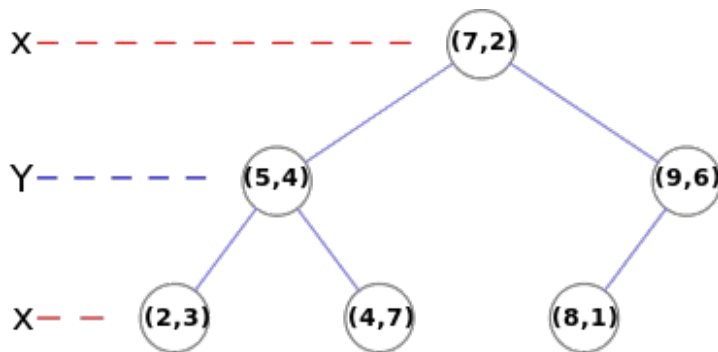
A popular practice is to sort a fixed number of randomly selected points, and use the median of those points to serve as the splitting plane. In practice, this technique often results in nicely balanced trees.

search



- Nearest neighbour search
- K nearest neighbours search
- Approximate nearest neighbour search
- Range search

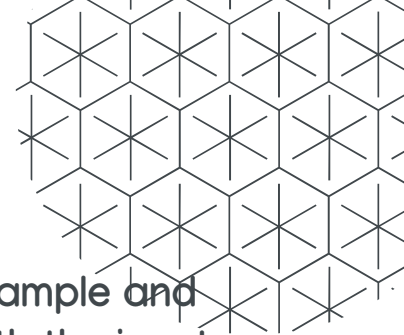
adding elements



- First, traverse the tree, starting from the root and moving to either the left or the right child depending on whether the point to be inserted is on the "left" or "right" side of the splitting plane.
- Once you get to the node under which the child should be located, add the new point as either the left or right child of the leaf node, again depending on which side of the node's splitting plane contains the new node.

Adding points in this manner can cause the tree to become unbalanced, leading to decreased tree performance

experiment



- Given 100.000 samples with `n_features` is 128. Then, input a any sample and convert it to vector 128. Let's find the nearest sample compare with the input.
- Solutions: linear search, k-d tree, etc.

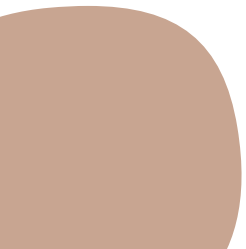
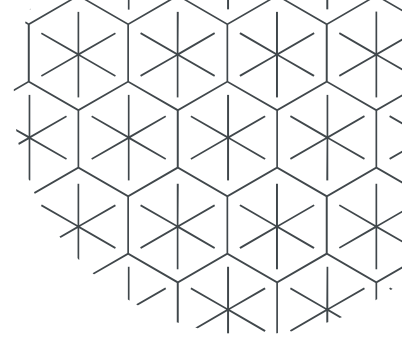
Solution	Time (s)	Accuracy(%)
linear search	0.44	100
k-d tree	0.02	100

Table comparison results between linear-search
and k-d tree

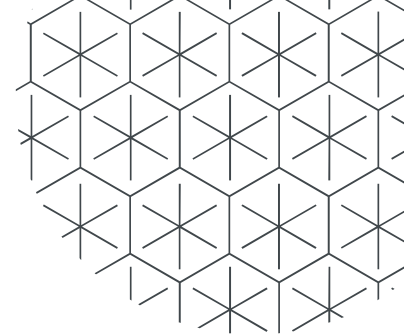
Algorithm	Average	Worst case
space	$O(n)$	$O(n)$
search	$O(\log n)$	$O(n)$
insert	$O(\log n)$	$O(n)$
delete	$O(\log n)$	$O(n)$

Time complexity k-d tree in big O notation

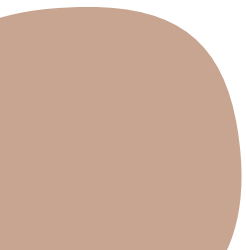
demo



homeworks



- Build k-d tree (build, nearest neighbour search, range search) from scratch
- Find out balancing, removing elements operators (optional)



The background features several abstract, organic shapes in muted colors: a large brownish-tan shape on the left, a large greyish-green shape on the right, and a smaller grey oval at the bottom right. A light beige, irregularly shaped area in the center serves as a backdrop for the text. Above the text, there is a simple black line drawing of a leafy branch.

**Thanks for
listening**