

UNIVERSITY NAME

DOCTORAL THESIS

Thesis Title

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*A thesis submitted in fulfilment of the requirements
for the degree of Doctor of Philosophy*

in the

Research Group Name
Department or School Name

February 2015

Declaration of Authorship

I, John SMITH, declare that this thesis titled, 'Thesis Title' and the work presented in it are my own. I confirm that:

- This work was done wholly or mainly while in candidature for a research degree at this University.
- Where any part of this thesis has previously been submitted for a degree or any other qualification at this University or any other institution, this has been clearly stated.
- Where I have consulted the published work of others, this is always clearly attributed.
- Where I have quoted from the work of others, the source is always given. With the exception of such quotations, this thesis is entirely my own work.
- I have acknowledged all main sources of help.
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Date:

“Thanks to my solid academic training, today I can write hundreds of words on virtually any topic without possessing a shred of information, which is how I got a good job in journalism.”

Dave Barry

UNIVERSITY NAME (IN BLOCK CAPITALS)

Abstract

Faculty Name

Department or School Name

Doctor of Philosophy

Thesis Title

by John SMITH

The Thesis Abstract is written here (and usually kept to just this page). The page is kept centered vertically so can expand into the blank space above the title too...

Acknowledgements

The acknowledgements and the people to thank go here, don't forget to include your project advisor...

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Abbreviations

LAH List Abbreviations **Here**

Physical Constants

Speed of Light $c = 2.997\,924\,58 \times 10^8 \text{ ms}^{-\text{s}}$ (exact)

Symbols

a	distance	m
P	power	W (Js^{-1})
ω	angular frequency	rads^{-1}

For/Dedicated to/To my...

Chapter 1

Background

1.1 Nearest Neighbors Search

1.2 Approximate Nearest Neighbors

1.3 k-d Trees

1.3.1 Overview

The k-d tree was originally developed as "a data structure storage of information to be retrieved by associative searches" [1]. k-d trees are efficient both in the speed of associative searches and in storage requirements.

A k-d tree is a binary tree which stores points in a k dimension space. Each node contains a single k-dimensional point, a split dimension, and up to two children nodes. Each node represents a hyperplane which lies perpendicular to the split dimension, and passes through the stored point. The left subtree of a node contains all points which lie to the left of the hyperplane, while the right subtree represents all points which lie to the right of the hyperplane. Thus, each node partitions all below it into two half-spaces. Because only a single split dimension is used, each splitting hyperplane is axis-aligned.

1.3.2 Construction

The construction of a k-d tree is performed recursively with input parameters of a list of points. Pseudo code is shown below in 0.

```
function KDTree(pointList)
```

```

splitDim = selectAxis()

medianPoint = selectMedian(pointList, splitDim)
leftList = select points less than medianPoint along splitDim
rightList = select points greater than medianPoint along splitDim

treenode node = new treenode()
node.splitDim = splitDim
node.splitPoint = medianPoint
node.leftChild = kdtree(leftList)
node.rightChild = kdtree(rightList)

return node
end function

```

Axis selection can be performed in multiple ways. The classical approach is to deterministically alternate between each dimension. Another approach known as spatial median splitting selects the the longest dimension present in the current pointList to split on [2]. The downside of this method is that a linear traversal is required to select the split dimension. Another popular approach is to randomly select the split dimension with an equal probability of selection each dimension. This approach is often applied when using multiple k-d trees as because of the additional randomness trees are more likely to be different.

While a linear time algorithm for determining the median of an unordered set is possible [3] a heuristic approach is typically used to approximate the median. A common heuristic is to take the median of five randomly chosen elements, however many other methods can be used such as the triplet adjust method [4].

At the termination of of 0, the root of the k-d tree is returned, and each node contains exactly one point. The runtime of this algorithm is $O(N \log(N))$ where N is the number of points in pointList. While the median can be approximated in constant time, partitioning pointList along that median is an $O(N)$ operation. Since the k-d tree is a binary tree, assuming it is relatively balanced, its height is $O(\log(N))$.

1.3.3 Nearest Neighbor Query

1.3.4 Modification

Appendix A

Appendix Title Here

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Bibliography

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