Big Data Lab, Jan. 31, 2015

## MENDEL: A DISTRIBUTED STORAGE SYSTEM FOR EFFICIENT SIMILARITY SEARCHES AND SEQUENCE ALIGNMENT

Jan. 30, 2015 Cameron Tolooee

#### Outline

- Motivation
- Overview
- Vantage-Point Tree
- System Architecture
- Results
- Conclusion & Future Work
- Questions

#### Motivation

- Due to exponential growth of biological datasets, current similarity search tools are becoming less sufficient
  - BLAST, BLAT, YASS, FASTA, etc...
  - Algorithm centric, different heuristics on similar algorithm with different trade-offs
- Similarity between sequences, or lack thereof, can explain relationships between them
  - In some cases can provide important clues about common evolutionary roots of organisms

#### Basic Idea

#### Inverted index

Map content to its location in the database

- Rather than indexing what each location contains
- Allows for efficient searches at the cost of additional processing for insertions
- DHTs provide extremely fast lookups for distributed datasets

#### Basic Idea

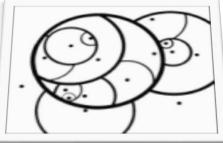
- Searching DNA sequences for subsequences is a challenging problem
  - Must consider partial matches, insertions/deletions (indels), repeated regions, etc..
- Sliding window over DNA sequence indexing on each substring
  - Sliding window can identify indels
- Store data with in a DHT with a nearest neighbor data structure
  - Nearest neighbor structure finds partial matches



- How to locate matches for a non-exact match query in a DHT?
- How to balance content load on storage nodes?

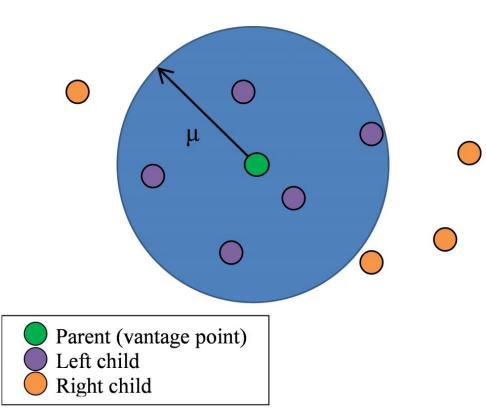
# Vantage Point Tree

- Developed by Peter Yianilos and Jeffrey Uhlmann independently
- Data structure used for nearest neighbor searches in metric space
- Recursively partition data points into two divisions
  - Points that are within a threshold distance of the vantage point
  - Points that are outside the same threshold

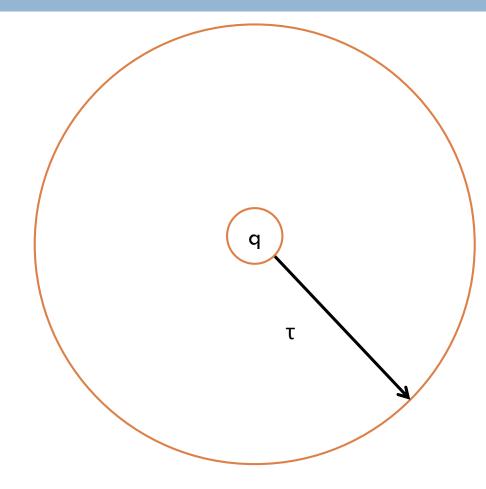


### Vantage-Point Tree

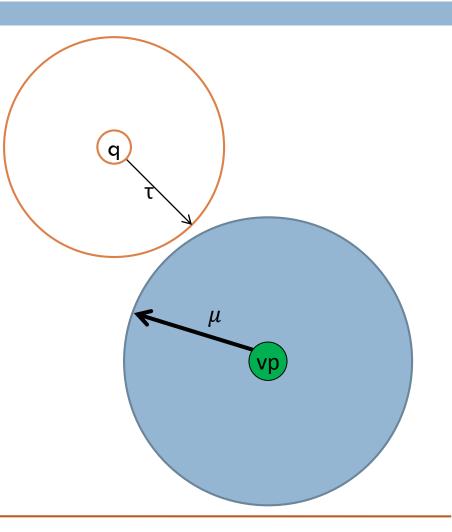
- Each node in a vp-tree maintains four values:
  - Input value
  - 🗖 Radius, µ
  - Left child
  - Right child



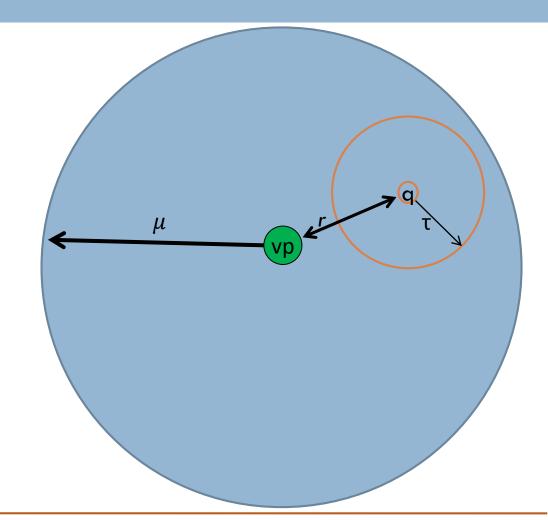
- Let query be q
- Let radius of q be  $\tau$
- k nearest neighbors are contained within τ
- $\tau = \min(dist(q \rightarrow v, \tau))$
- 3 cases
  - $\tau$  lies completely within  $\mu$
  - $\tau$  lies completely outside  $\mu$
  - $\tau$  and  $\mu$  intersect
- Stop recursing when leaves are reached



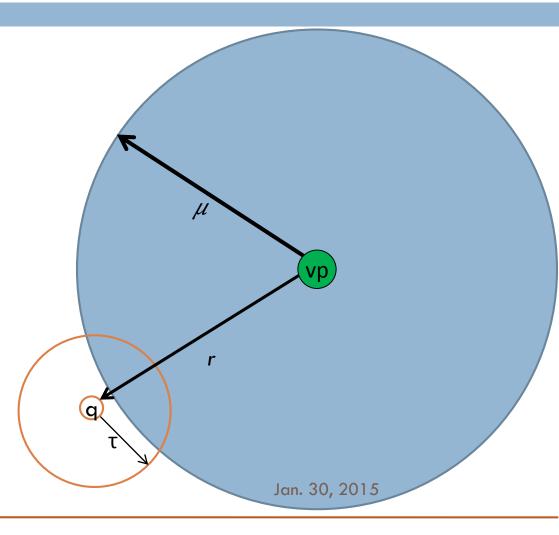
- Case 1:  $\tau$  completely outside of  $\mu$ 
  - Don't need to search left subtree
- $\tau = \min(dist(q \rightarrow vp), \tau)$ 
  - Recurse on right subtree



- Case 2: τ completely inside μ
  - Don't need to search right subtree
- $\tau = \min(r, \tau)$ 
  - Recurse on left subtree

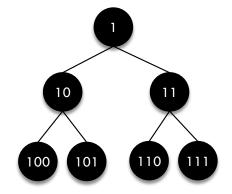


- Case 3: worst case intersect
  - Must search both trees
- $\tau = \min(r, \tau)$ 
  - Recurse on both subtrees



#### Vantage-point prefix tree

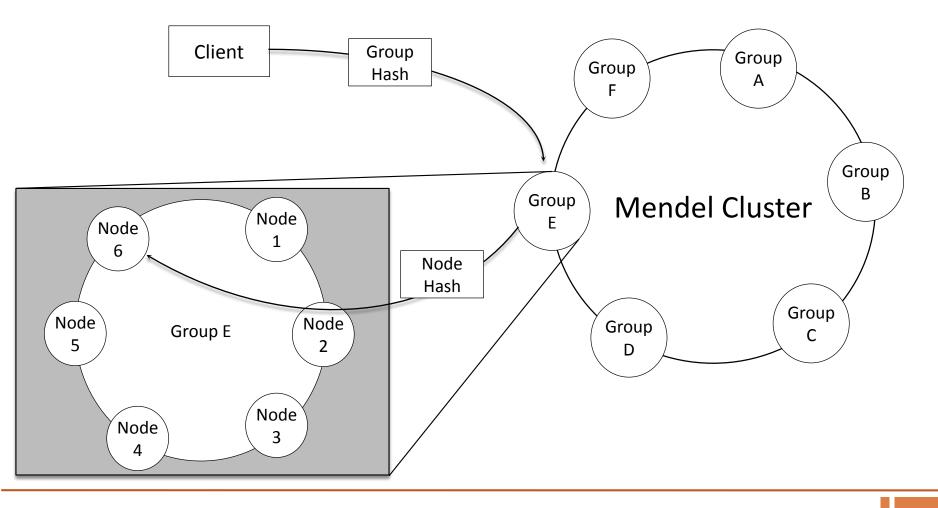
- Global vp-tree as an index is not scalable
  - Utilize vp-tree as a similarity based hashing function
- Alter vp-tree node to contain a prefix
  - $prefix_{left} = prefix_{parent} \ll 1$  $prefix_{right} = (prefix_{parent} \ll 1) + 1$
- Use as a group hash by assigning groups to subtrees
  - Requires a balanced vpp-tree



### System Architecture

- Zero-hop distributed hash table
  - Such as Apache Cassandra and Amazon Dynamo
- Hierarchical, two-tier hashing scheme
- Each node belongs to a group
  - Groups are placed on the hashing ring
  - Two rounds of hashing required to place or retrieve data
    - Hashed to a group using the vpp-tree
    - Second hash among group nodes

### System Architecture



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## Indexing Data

- 100bp sliding window over each contig
  - Each 100bp subsequence is individually indexed
- Passed through the vpp-tree to determine storage grouping
- Within the group, the subsequence is distributed using a SHA-1 hash to a storage node
- The subsequence block is maintained in a vp-tree local to its storage node

## **Query Evaluation**

- Query is "hashed" in the vpp-prefix tree to find all subtrees that may have matching subsequences
- Each node in the selected group(s) performs a lookup in their vp-tree
  - Results are aggregated and filtered
- Results are send back to the client

#### Results

- Three benchmarks to test indexing speed, data distribution, and query speed
- Sourced real world data from the Genome Assembly Golden-standard Evaluation (GAGE)

Four genomes ranging from 2 Mbp to 3 Gbp

Benchmark 1: index each of the genomes into the system and measure the time to complete

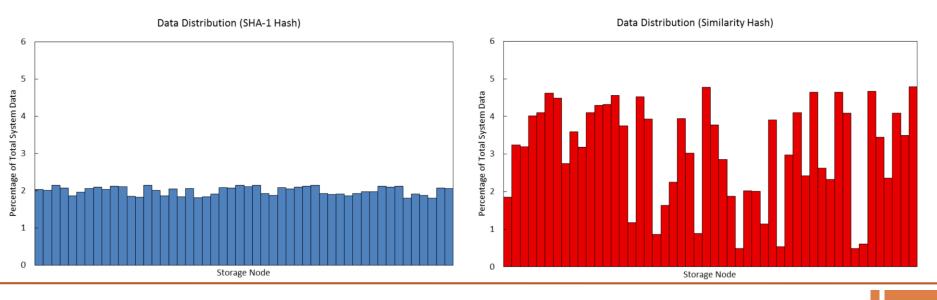


#### TABLE I INDEXING TIMES

Genome	<b>Base Pairs</b>	Blocks	Index Time
S. aureus	2.8 Mbp	28,261	1.80 s
R. sphaeroides	4.6 Mbp	45,984	2.61 s
H. sapies C. 14	88 Mbp	882,468	21.83 s
B. impatiens	250 Mbp	2,491,627	88.14 s

#### Results

- Benchmark 2: Data distribution
  - After all datasets have been indexed count files per node
  - Compare versus flat SHA-1 hash



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#### Results

- Benchmark 3: Issue a series of queries; measure response time and number of results
  - Exact match query whose target exists in the database
  - Exact match query whose target has a few errors to its match
  - Similarity query whose target exists in the database
  - Similarity whose target has a few errors to its match
  - Similarity whose target is randomly generated



#### TABLE II Retrieval Times

Query	Number of results	Time (ms)
Exact Match, exists	1	403
Exact Match, erroneous	0	346
Similarity, exists	8	409
Similarity, erroneous	8	476
Similarity, random	10	480

# **Conclusion & Future Work**

- The hashing scheme needs to be refined substantially in order to level the out the dispersion of the data
  - Data input one-by-one
  - Choosing initial vantage point (root)
- Currently queries must match the window they were indexed with
  - Sliding window over queries



#### Thanks!