Analysis and performance optimization of k - Nearest Neighbor approach for movie queries

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The Idea

- To implement a movie query engine that allows users to find a movie of interest by starting the query from a movie that they have already seen
- Example of a query –

"When Harry met Sally" - Romance + Thriller = "From dusk till dawn"

Data and Representation

- Custom collected data set crawled from
 - Freebase
 - MovieLens 10M
- Feature Vector
 - 3484 x 276 vector (row = movies, column = genres)
 - Each genre has a value o or 1
 - o if movie is categorized in that genre
 - 1 if it isn't

ML Algorithm

- kNN star of the show
- Optimization of KNN
 - Pre-clustering using k-means
 - kd-trees
 - Space partitioning data structure for organizing points in a k-dimensional space
 - Locality Sensitivity Hashing(LSH)
 - method of performing probabilistic dimension reduction of high-dimensional data

K-d Trees vs. LSH

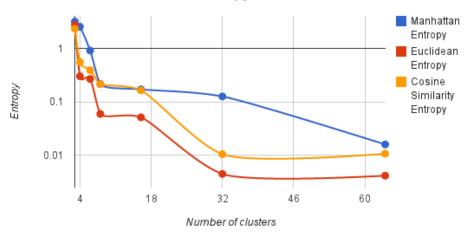
- If the dimensionality of the data is k, a k-d tree works well only if the number of data points is more than 2^k.
- Otherwise, we end up testing nearly all the nodes in the dataset making the complexity O(n) instead of the desired O(log(n)).
- LSH is expected to perform much better when the number of dimensions is large, like in our case.

Results

- Tasks Completed:
 - kNN classification
 - K Means clustering of the data set using various distance metrics and values of k
 - kNN Classification on clustered data
- Tasks Remaining:
 - Comparison with k-d trees or LSH

Results



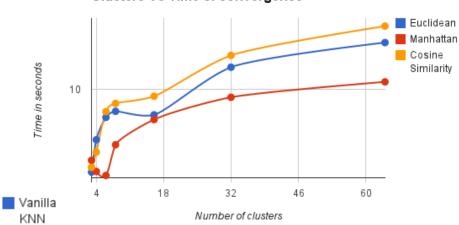


Clusters Vs Time of convergence

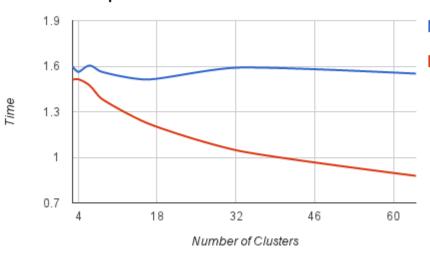
Time in seconds

KNN Clustered

KNN



Comparison - KNN vs Clustered KNN



Project Takeaways

• For us:

- MQL (Metaweb Query Language)/Freebase A new API and data repository
- Ability to assess trade-offs and choose appropriate optimization methods based on dataset in hand (i.e. understanding of scenarios where certain algorithms fail)
- An in-depth understanding of KNN

For you:

 A new movie query engine (Yes, we plan to put it online with an interactive interface)

References

- Locality-Sensitive Hashing for Finding Nearest Neighbors [Malcolm Slaney and Michael Casey]
- Multidimensional Binary Search Trees Used for Associative Searching [Jon Louis Bentley, Stanford University]
- Our best friend, <u>www.google.com</u>

Thank You

Questions?