

A Distributed Greedy Heuristic for Computing Voronoi Tessellations With Applications Towards Peer-to-Peer Networks

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Motivation

We were creating a distributed network based off of Delaunay Triangulation and needed a fast distributed algorithm for calculating Delaunay peers. We ran into a couple of issues.

- ▶ Distributed algorithms for solving Delaunay triangulation don't really exist (every node does a global solution).
- ▶ Not any fast solutions if we start moving out of 2D Euclidean space.

This leaves approximation. However:

- ▶ Simple approximation doesn't guarantee a fully-reachable network (k -nearest neighbor, nodes in radius r).
- ▶ Other solutions [1] required prohibitively high sampling or other hidden time costs.

And nothing can handle moving nodes.

Voronoi Tessellation and Delaunay Triangulation

- ▶ Voronoi Tessellation
 - ▶ Partition space such that each point is mapped the 'voronoi generator' to which it is closest.
- ▶ Delaunay Triangulation
 - ▶ generate a triangulation, such that, for every triangle the circumcircle does not contain any other points

Voronoi Example

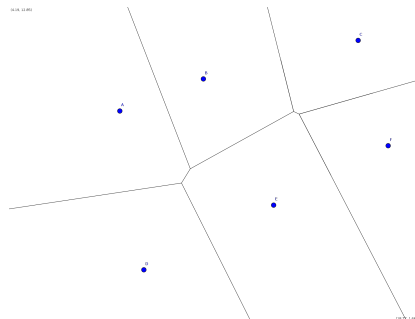
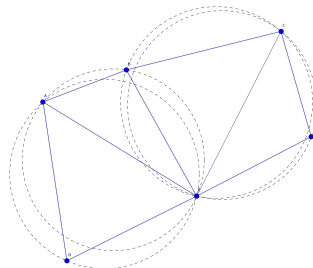


Figure: The space is partitioned, such that each point in a partition is closest to that partitions 'Voronoi generator'

Delaunay Example

0.00, 0.00



100.00, 1.00

Figure: The space is partitioned, such that each point in a partition is closest to that partitions 'Voronoi generator'

Distributed Hash Tables

- ▶ Abstractly, a DHT is a mechanism for maintaining a large state in a decentralized network.
- ▶ In practice, the state is a large number of (*key*, *value*) records.
- ▶ A Distributed hash table assigns those records to servers and routes request for those records to those servers.
- ▶ Current incarnations of Distributed hash tables assign servers and records locations in an arbitrary metric space.
- ▶ Servers are assigned responsibility for records that are “close” to them, and to peer with “nearby” servers.
- ▶ DHTs currently use a variety metric spaces.

How are DHTs and Voronoi Tessellation/Delunay Triangulation related?

A Server is responsible for records "close" to it (Voronoi Triangulation). A Server peers with other servers that bound it's Voronoi Region (Delunay Triangulation). DHTs often have peers in excess of the Delaunay triangulation to shorten lookups.

- ▶ Ring Based DHTs
 - ▶ Chord: a unidirectional modulus ring metric
 - ▶ Pastry, Symphony: bidirectional modulus ring
- ▶ Tree Based DHTs
 - ▶ CAN: Euclidean distance
 - ▶ Kademlia: XOR distance

Applications of DHTs

- ▶ *P2P file sharing* is by far the most prominent use of DHTs. The most well-known application is BitTorrent [4].
- ▶ *Distributed Domain Name Systems* (DNS) have been built upon DHTs [5] [7]. Distributed DNSs are much more robust than DNS to orchestrated attacks, but otherwise require more overhead.
- ▶ Distributed *machine learning* [6].
- ▶ Many *botnets* are now P2P based and built using well established DHTs [8]. This is because the decentralized nature of P2P systems means there's no single vulnerable location in the botnet.

Why do we need a distributed Voronoi heuristic?

- ▶ The different topologies DHTs utilize present optimization trade-offs (lookup latency, number of lookup hops, network robustness, availability, processing overhead)
- ▶ The primary effort in implementing a new metric space in a DHT is implementing Voronoi Regions/Delunay Triangulation algorithm in that metric.
- ▶ DGVH allows many metrics to be tested without requiring the design and development effort of generating an exact Voronoi Regions/Delunay Triangulation algorithm.

So we created a Distributed Greedy Voronoi Heuristic, or DGVH for short.

Distributed Greedy Voronoi Heuristic

- ▶ Geometrically intuitive method of approximating the one-hop delaunay peers of a Node
- ▶ Is guaranteed to form a connected mesh (unlike k-nearest heuristic)
- ▶ Can be utilized in any continuous metric space

DGVH Intuition

- ▶ The majority of Delunay links cross the corresponding Voronoi edges.
- ▶ We can test if the midpoint between two potentially connecting nodes is on the edge of the voronoi region.
- ▶ This intuition fails if the midpoint between two nodes does not fall on their voronoi edge.

DGVH Algorithm

- 1: Given node n and its list of *candidates*.
- 2: $peers \leftarrow$ empty set that will contain n 's one-hop peers
- 3: Sort *candidates* in ascending order by each node's distance to n
- 4: Remove the first member of *candidates* and add it to *peers*
- 5: **for all** c in *candidates* **do**
- 6: m is the midpoint between n and c
- 7: **if** Any node in *peers* is closer to m than n **then**
- 8: Reject c as a peer
- 9: **else**
- 10: Remove c from *candidates*
- 11: Add c to *peers*
- 12: **end if**
- 13: **end for**

DGVH Example

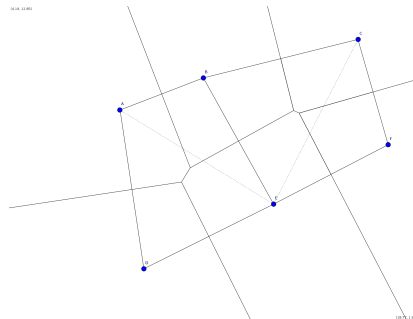


Figure: 'DGVH connects all nodes where the midpoint falls on the border of the voronoi regions'

Realistic Candidate set Size

- ▶ In application, a single node will not need to calculate the triangulation for every peer in the network, rather it will only calculate it's own peers.
- ▶ A smart "join" process will prevent the Candidate Set from reaching $O(n)$
- ▶ Expected size of the candidate set is $O(\text{degree}^2)$ which compromises current peers and 2-hop peers
- ▶ The average degree of many metric spaces is $O(1)$, then therefore DGVH will practically run in $O(1)$ time in those metric spaces.

Error Mitigation

- ▶ As DGVH is a heuristic, it will likely have errors when compared to a global Delaunay triangulation.
- ▶ This error rate is difficult to find analytically and will vary between metric spaces and distributions of locations
- ▶ A simple method of mitigating this error is to keep both one and two hop peers.

Experiment 1

In our first experiment, we wanted to test the accuracy of our heuristic.

- ▶ We generated a random graph and created the Delaunay triangulation using a global solution and DGVH.
- ▶ We found DGVH had approximately one error per node.

Our second set of experiments demonstrate that this error rate is sufficiently low for distributed applications.

Results

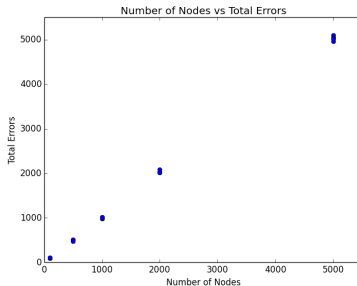


Figure: As the size of the graph increases, we see approximately 1 error per node. We can also see that the error rate and number of nodes has a linear relationship.

Experiment 2

Our second experiment were designed to test how well nodes could form a routing topology using DGVH.

- ▶ For each trial, we create a random graph.
- ▶ During the first two cycles, nodes are given 10 random connections.
- ▶ In each cycle (including the first two):
 - ▶ Each node gossips and runs DGVH.
 - ▶ 2000 random lookups from random nodes to random locations performed.
- ▶ The rate of successful lookups approaches 1.0 as time progresses.

Results

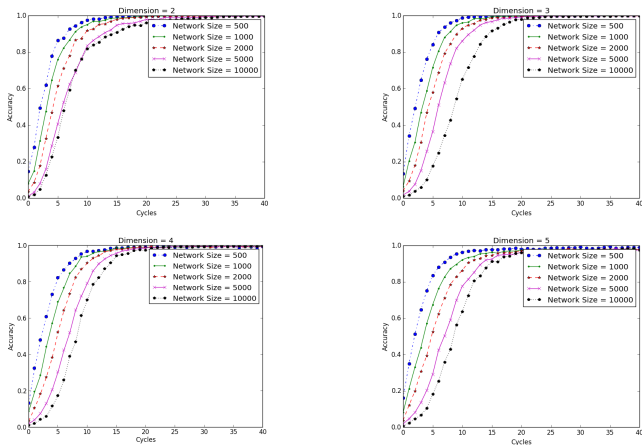


Figure: These figures show, starting from a randomized network, DGVH forms a stable and consistent network topology. The Y axis is the percentage of successful lookups out of 2000 queries and the X axis is the number of gossip cycles.

Other Applications

One type of distributed system that can use Voronoi tessellations are *wireless ad-hoc networks*.

- ▶ Solves the coverage-boundary problem [2].
- ▶ Can be used for sleep scheduling [3]

Conclusions

- ▶ DGVH provides a simple approximation Voronoi Triangulation / Delaunay Triangulation, of similar complexity to picking k-nearest node or nodes in distance k.
- ▶ Unlike other simple approximation methods, DGVH guarantees a fully connected graph.
- ▶ In practice, DGVH supplemented with K-nearest peers approximation provides an accurate, fast and distributable computation, with a fully connected graph as result.



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