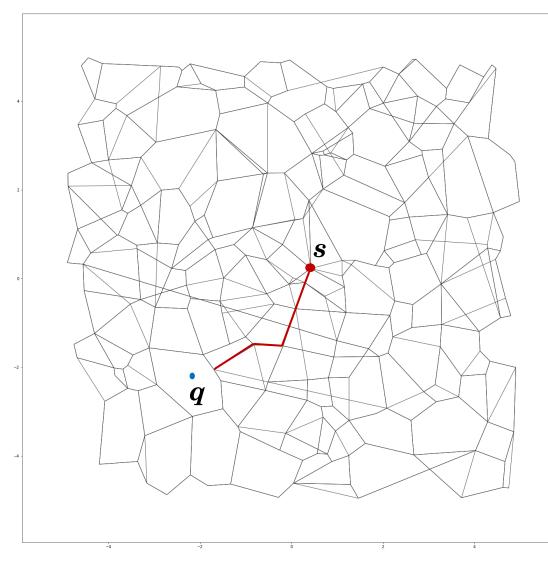
## Carnegie Mellon University

## Introduction

## K Nearest Neighbor search problem Embed cute cat photo" Embed Query Database K=4 nearest $N = 10^{6}$ to $10^{9}$ points dimension d~100-1000 neighbors $(I_2)$

# **Current Solution**

## Graph based ANN indices



Index: One vertex per data point/embedding. Directed edges between vertices.

### For query *q*:

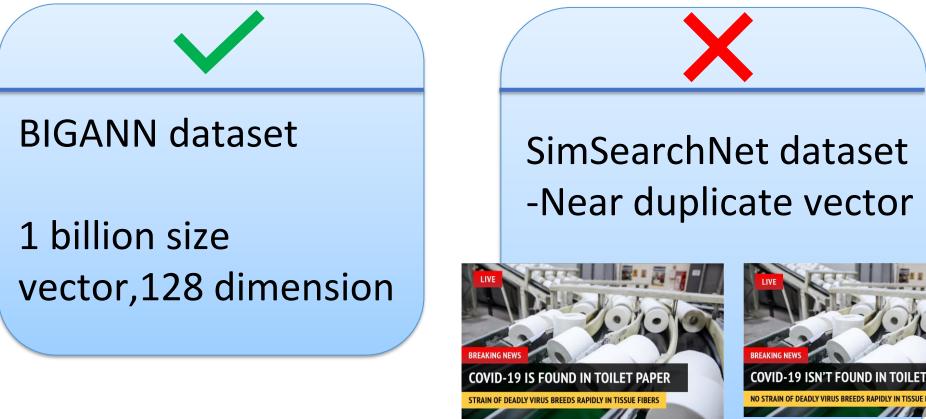
Start at point *s*, and iterate: 1. compute dist. from **q** to

neighbors 2. hop to node closest to **q**, as long as distance improves

Index build: Create a lowdegree graph that guides queries with the **fewest** hops and distance comparisons

Starting from one vertex per data

**Problem**: Graph Based algorithm is the fastest solution, but performs bad on specific dataset



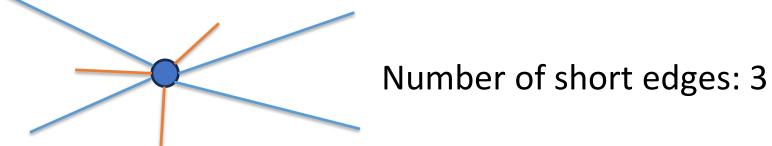


# Investigation on Sim search data set via ParlayANN library Taekseung Kim, Guy Blelloch, Magdalen Dobson (Carnegie Mellon University)

# Methodology

Hypothesis 1: Near-duplicates cause cluster-like structures in the graph and thus prevent the greedy search from converging

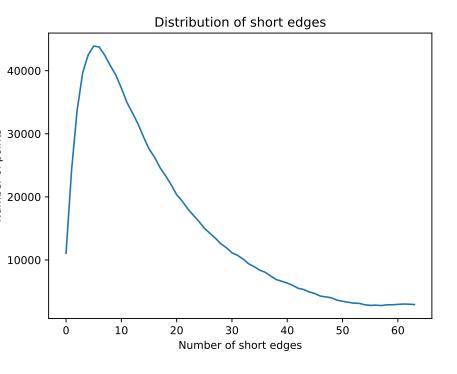
1. Found the **distribution of the short edges** of generated graph in SimSearchNet dataset. Short edges: Compared diameter of a dataset. With more short edges, there likely is a cluster.



2. Check the **distance distribution** of SimSearchNet dataset, compared to BIGANN dataset Hypothesis 2: Starting from **multiple points** would make the algorithm faster. 3. Implement a new method of search. Start from two starting points, using the second starting point as the true closest neighbor.

## Results

## **Experiment 1: Distribution of short edges**

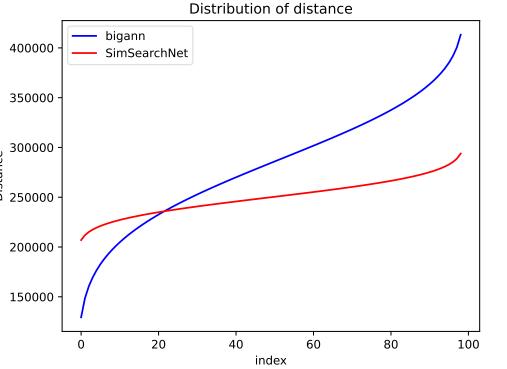


We found there are not a lot of short edges among the vertices

Rather, most of the points have 0~6 short edges, where their maximum degree is 64

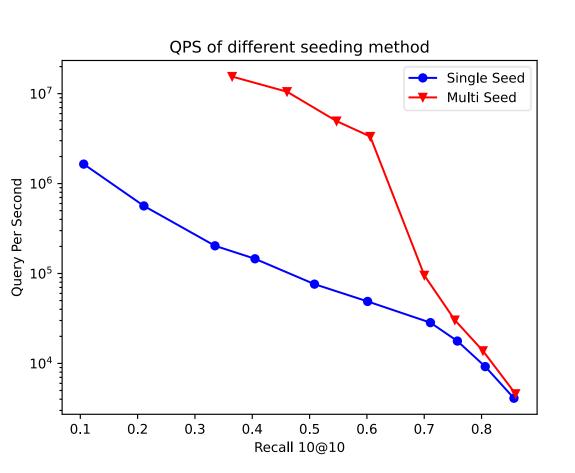
This fact implies there might not be a lot of clusters

**Experiment 2: Distance distribution** of the SimSearchNet dataset, compared to BIGANN dataset



SimSearchNet dataset has a narrower distance distribution, which means the distance among points is similar. This prevents the greedy search from converging.

**Experiment 3: QPS(Query Per Second)** comparison of multiple and single starting point

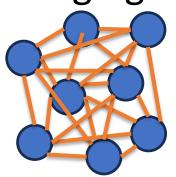


We can see that the multi seed method is much faster at finding 0.6 recall(which is 6 correct points out of 10), but gets almost the same throughput(Although it is about 1.4 times faster)

## Discussion

### Hypothesis 1:

- Number of Short edges: There are not a lot of **clusters** in SimSearchNet dataset, since there is small amount of short edges
- **Distance distribution**: Compared to the BIGANN dataset, SimSearchNet dataset has a **narrow distribution**. A narrow distribution is likely to prevent the greedy search from converging.



One cluster induces vast amount of short edges

### Hypothesis 2:

• **Seeding method:** Searching from multiple points, we found that by providing one nearest neighbor we get 4 of the nearest points immediately. However, the **overall speed** for the 10 nearest neighbors has x1.4 speedup, which is not as much as we expected.

# **Future Work**

- We plan to try more **optimizations** on searching and try some other searching algorithms.
- Other prospective methods: Bucket-based methods can be integrated into the search algorithm to speed up search for the SimSearchNet dataset.





3 buckets, based on distance of source