How to select hashing bits with a simple measurement? An alternative greedy approach

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Hashing algorithms for KNN problems

- Locality Sensitive Hashing (LSH) Preserve the similarity if enough bits are used.
- Spectral Hashing Design compact bits which preserves the local similarity.
- Iterative Quantization (ITQ) Minimize the quantization error.

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Previous bits selection methods



Motivation



Measuring the quality of a bit set

For each query data point, its nearest n data points in sense of the Hamming distance are retrived. Recall@n is the percentage of true nearest neighbor points in the retrieved data set, i.e.:

$$\operatorname{Recall}@n = \frac{\#\operatorname{retrieved true nearest neighbor points}}{\#\operatorname{true nearest neighbor points}}$$

Then the m-Recall is calculated as:

m-Recall =
$$\frac{\sum_{n=1}^{K} \text{Recall}@n}{K}$$

here K is the maximum retrieved nearest points in Hamming space and Recall@n is the average recall of retrieving n nearest neighbor points.

Bits selection with alternative greedy method I

- ► The task is to select a set S with M bits from the pool P with size N.
- S is initialed with the bits randomly selected from P.
- ► For updating the *i*-th bit, the position is updated with the bit which has the highest m-Recall score.

Bits selection with alternative greedy method II

- For updating the *i*-th bit, the hamming distance between the remaining M 1 bits can be precomputed.
- Measuring each candidate bit parallelly.
- The optimization method is guaranteed to obtain a local optimal solution.

Why it works?



Results



NUSWIDE, 128D, training (10,000), query (10,000), base datasets (249,648), Pool (500).

- Retrieving task: 32 bits, 1-NN.
- Performance metric:

 $Recall@N = \frac{\text{The number of retrieved true nearest neighbor points}}{\text{The number of retrieved data points}}$

Thank you for your time!