



An Alternating Optimization Scheme for Binary Sketches for Cosine Similarity Search

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Binary Sketching & Indexing

- Binary sketching defines a map $H: X \rightarrow \{0, 1\}^B$
- "Quality" of sketches is induced by downstream applications
- For indexing, quality of sketches often

 $Q(H) \approx \operatorname{Corr}_{x,y \in X} \left(d_{Hamming}(H(x), H(y)), d_X(x, y) \right)$

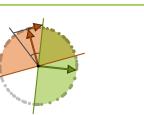
- Approximate *k*-nn search by, e.g.:
 - Scan H(X) for k'-nn with d_{Hamming} (cheap; many)
 - Refine k'-nn with d_x to k-nn (expensive; few)

The (Euclidean) Spherical Case

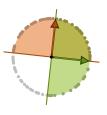
- Most natural separation by dot product
 ⇒ B hyperplanes one per bit
 ⇒ Tessellation of the d-sphere
- "Optimal" tossol, should have homed
- "Optimal" tessel. should have homogeneous sample counts, surface density integrals, and shapes
- "Balance" (∝ "entropy") of bits can be maximized without affineness

Alternating Optimization (HIOB)

- Idea: Improve initial hyperplanes by rotation
 - Homogeneous sample counts induced by pairwise independent bits of hyperplanes
 - Hope for surface area and shape to "work out" (by adding noise to X)
- Rotation by additive tangential vector (see Figure 1)
 ⇒ Aggregation of multiple updates if desired
- Scale rotation angle to help with convergence
- Work on varying subsamples to speed process up
- Observation: With "good" initialization, always only updating "worst offenders" works best



(a) Before update



(b) After update

Figure 1: Example of an update with exactly two planes.

Evaluation

- Our approach improves bit "balance" (see Figure 2)
- "Indexing quality" of sketches improves aswell but can fall off (see Figure 2)
- Bruteforce search on optimized sketches can outperform some indices (see challenge)
- HNSW is still much faster, but builds much slower

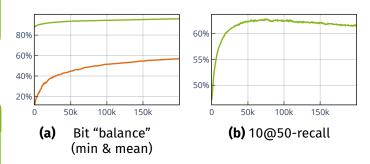


Figure 2: Balance and recall over iterations of HIOB

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- Intrinsic Dimensionality
- High-dimensional data



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