Approximate Nearest Neighbors in Hamming Model

Algorithmic Problems Around the Web #6

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Self-Reduction in a Nutshell

Problem: $(1 + \varepsilon)$ -approximate *I*-range queries in *d*-dimensional Hamming cube

Self-Reduction in a Nutshell

Problem: $(1 + \varepsilon)$ -approximate I-range queries in d-dimensional Hamming cube

- Apply embedding $\{0,1\}^d$ into $\{0,1\}^k$ such that I-neighbors usually fall within $\delta_1 k$ from each other, while $(1+\varepsilon)I$ -far objects are embedded at least $\delta_2 k$ from each other
- Precompute all $(\frac{\delta_1+\delta_2}{2})k$ -neighbors for every point in $\{0,1\}^k$
- In search step, embed q and explicitly check all precomputed $(\frac{\delta_1+\delta_2}{2})k$ -neighbors

RP: Inner product test

Single test:

- Choose random subset of positions of size $\frac{1}{2l}$
- Randomly assign 0 or 1 to every of them, the rest assign to 0, call the resulting vector r
- $\bullet \ h_r(p) = r \cdot p$

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Claim: there exist constants $\delta_1 > \delta_2$

- $H_d(p,s) \leq I \Rightarrow Pr[h(p) = h(q)] \geq \delta_1$
- $H_d(p,s) \ge (1+\varepsilon)I \Rightarrow Pr[h(p)=h(q)] \le \delta_2$

RP: Preprocessing

Inner product mapping:

- Choose k random tests r_1, \ldots, r_k
- Map every p into $A(p) = h_{r_1}(p) \dots h_{r_k}(p)$

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Data Structure

- Apply inner product mapping to all strings in database
- ullet For every $v\in\{0,1\}^k$ precompute all $(rac{\delta_1+\delta_2}{2})k$ -neighbors

RP: Search

- Compute $A(q) = h_{r_1}(q) \dots h_{r_k}(q)$
- Retrieve and explicitly check all $(\frac{\delta_1+\delta_2}{2})k$ -neighbors of A(q)

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Analysis:

- Chances to miss true *I*-neighbor: $\exp(-\frac{\delta_1 \delta_2}{2\delta_1}k)$
- Chances to waste time on $(1+\varepsilon)$ /-far neighbor: $\exp(-\frac{\delta_1-\delta_2}{2\delta_1}k)$

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Thus we should take near-logarithmic k which lead to polynomial size of $\{0,1\}^k$ to be NN-precomputed

RP: Formal Claim

Theorem (Kushilevetz, Ostrovsky, Rabani, 1998)

Consider $(1+\varepsilon)$ -approximate I-range search in d-dimensional Hamming cube. Then for every μ there is a randomized algorithm with (roughly) d^2 polylog(d, n) query time and $n^{\mathcal{O}(\varepsilon^{-2})}$ preprocessing space. For every query this algorithm answers correctly with probability at least $1-\mu$

Thanks for your attention! Questions?

References

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http://simsearch.yury.name/tutorial.html



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