

# 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  $l$ -range queries in  $d$ -dimensional Hamming cube

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

- Apply embedding  $\{0, 1\}^d$  into  $\{0, 1\}^k$  such that  $l$ -neighbors usually fall within  $\delta_1 k$  from each other, while  $(1 + \varepsilon)l$ -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  $r$
- $h_r(p) = r \cdot p$

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

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

# RP: Preprocessing

## Inner product mapping:

- Choose  $k$  random tests  $r_1, \dots, r_k$
- 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
- For every  $v \in \{0, 1\}^k$  precompute all  $(\frac{\delta_1 + \delta_2}{2})k$ -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  $l$ -neighbor:  $\exp(-\frac{\delta_1 - \delta_2}{2\delta_1} k)$
- Chances to waste time on  $(1 + \varepsilon)l$ -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 (Kushilevets, Ostrovsky, Rabani, 1998)

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

Thanks for your attention! Questions?

# References

**Course homepage**      <http://simsearch.yury.name/tutorial.html>



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