Weekplan: Approximate Near Neighbor

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References and Reading

[1] Notes by Aleksandar Nikolov

[2] Near-optimal hashing algorithms for approximate nearest neighbor in high dimensions, Andoni and Indyk, Communications of the ACM, January 2008.

We recommend reading [1] in detail and [2] section 1-3.

Probability theory cheat-sheet

Markov's inequality: For *Y* being a positive-valued random variable,

$$P[Y \ge t] \le \frac{\mathbb{E}[Y]}{t}$$
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Exercises

1 Hamming Distance Let P be a set of n bit vectors each of length d. Give a data structure (for Hamming Distance) for which INSERT can be implemented in O(1) time and NEARESTNEIGHBOR in O(nd) time.

2 LSH Hamming Distance Let T_1 , T_2 be hash tables of size 5 with hash functions $h_1(x) = (3x + 4) \mod 5$ and $h_2(x) = (7x + 2) \mod 5$. Let $g_1(x) = x_1x_4x_8$ and $g_2(x) = x_1x_7x_7$.

Insert the bit strings x = 10110011, y = 10001101, z = 00110010, u = 01001010, v = 01001000 and draw the hash tables. Compute the result of ApxNearNeighbor(10111010).

3 *c*-Approximate Closest Pair under Hamming Distance Assume you have a data structure bit vectors for which INSERT and APXNEARESTNEIGHBOR run in time T(n). The distance metric is the Hamming distance, n is the number of bit strings in the data structure, and APXNEARESTNEIGHBOR(x) returns a point no more than $c \cdot \min_{x \in P} d(x, z)$ away from x.

Give an algorithm that given a set P of n bit vectors each of length d finds a pair x, y of distinct strings in P in time $O(T(n) \cdot n + dn)$ such that $d(x, y) \le c \cdot \min_{u,v \in P, u \ne v} d(u, v)$.

4 Hamming Distance Analysis From the proof of Claim 1 on the slides: Prove that $Lp_1^k \ge 2$. *Hint*: Recall that $k = \lg n / \lg(1/p_2)$.

5 Hamming Distance Analysis **2** In this exercise we will analyse the LSH scheme for Hamming Distance. Recall that in a query we stop after we have checked 6L+1 strings. Let $F = \{y \in P : d(x,y) > cr\}$ (strings far from x) and let z^* be a fixed string with $d(x,z^*) \le r$. We say that y collides with x if $g_j(x) = g_j(y)$ for some $i \in \{1,\ldots,L\}$.

5.1 Explain why it is enough to prove that the following two properties hold:

- 1. the number of strings in F that collides with x at most 6L.
- 2. z^* collides with x.

5.2 Let y be a string in F. Prove that $P[y \text{ collides with } x \text{ in } T_i] \leq 1/n$. Hint: Recall that $k = \log n / \log(1/p_2)$.

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- **5.3** Let $X_{y,j} = 1$ if y collides with x in T_j and 0 otherwise, and let $X = \sum_{y \in F} \sum_{j=1}^{L} X_{y,j}$. Prove that $E[X] \le L$.
- **5.4** Use Markov's inequality to show that P[X > 6L] < 1/6.
- **5.5** Prove that if there exists a string z^* in P with $d(x,z^*) \le r$ then with probability at least 2/3 we will return some y in P for which $d(x,y) \le cr$.
- **6** Jaccard distance and Sim Hash The *Jaccard similarity* of two sets is defined as $JSIM(A, B) = \frac{|A \cap B|}{|A \cup B|}$. In *Min-Hash* you pick a random permutation π of the elements in the universe and let $h(A) = \min_{a \in A} \pi(a)$.
 - **6.1** Let $S_1 = \{a, e\}, S_2 = \{b\}, S_3 = \{a, c, e\}, S_4 = \{b, d, e\}$. Compute the Jaccard similarity of each pair of sets.
- **6.2** Let S_1, S_2, S_3, S_4 be as above and let the random permutation be (b, d, e, a, c), i.e., $\pi(a) = 4$, $\pi(b) = 1$, etc. Compute the min-hash value of each of the sets.
- **6.3** Prove that the probability that the min-hash of two sets is the same is equal to the Jaccard similarity of the two sets, i.e., that $P[h(A) = h(B)] = \frac{|A \cap B|}{|A \cup B|}$.
- **6.4** The Jaccard distance is defined as $d_J(a, b) = 1 JSIM(A, B)$. Show that the Jaccard distance is a metric. That is, show that:
 - 1. $d_J(A, B) \ge 0$ for all sets A and B,
 - 2. $d_I(A, B) = 0$ if and only if A = B,
 - 3. $d_{J}(A,B) = d_{J}(B,A)$,
 - 4. $d_J(A, B) \le d_J(A, C) + d_J(C, B)$ for all sets A, B and C.

Hint: For 4. use $P[h(A) = h(B)] = \frac{|A \cap B|}{|A \cup B|}$.