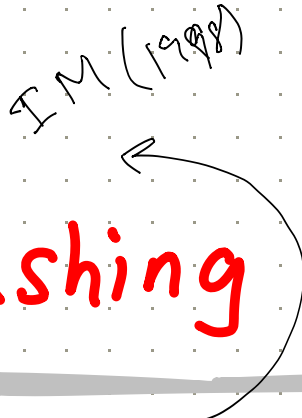


Approximate Nearest Neighbor
using

Locality Sensitive Hashing



IM (1998)

Based on Lecture Notes by Sarel Har-Peled

Presented by

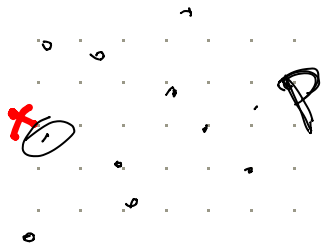
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The Nearest Neighbor Problem

→ Input: Set of points P in a metric Space (X, dist)

→ Goal: Preprocess them such that the following queries can be efficiently solved.

Query: Given a point $q \in X$, output the point $p \in P$ such that $\text{dist}(p, q)$ is minimized.



The Nearest Neighbor Problem

Tradeoff between Preprocessing Space and Query time.

One extreme:-

Don't preprocess (constant preprocessing space)

$|P| = n$ query time

Other extreme:

Compute Voronoi Diagram ($n^{d/2}$ space required)

Constant Query time.

Approximate Nearest Neighbor (ANN)

Input: A set of points P in a metric space (X, dist)

Goal: Preprocess set P s.t. for query $q \in X$, return $p \in P$

$$\text{dist}(q, p) \leq (1 + \epsilon) \text{dist}(q, p^*)$$

where

p^* = nearest neighbor of q in $P =: \text{nn}(q)$.

Approximate Nearest Neighbor (ANN)

We will assume $\mathcal{X} = \mathbb{R}^d$.

Naive method: nd query time

Goal: $\text{Sub}(n) \cdot d$ query time with reasonable space

This Presentation:

Query time: $\tilde{O}\left(d n^{\frac{1}{1+\epsilon}}\right)$ whp

Preprocessing space: $\tilde{O}\left(nd + n^{\frac{1}{1+\epsilon}}\right)$

A Simpler Problem: Proximity Neighbor (PN)

$PN(q, r)$:

For a given query point q :

- 1) If there is $p \in \text{Ball}(q, r)$, return it.
- 2) If no point in $\text{Ball}(q, r)$, say so.



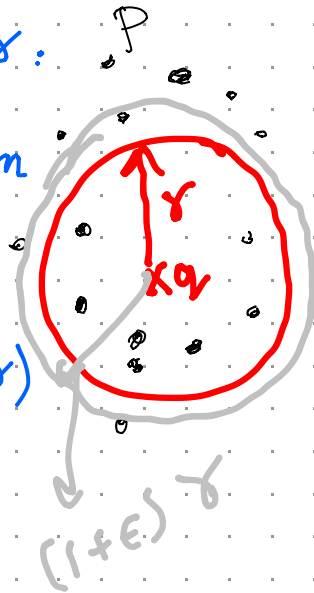
Approximate Proximity Neighbor (APN)

Fix accuracy parameter ϵ .

APN(q, r):

For a query point q , and a radius r :

- 1) If there is a point in $\text{Ball}(q, r)$, return
Some point in $\text{Ball}(q, (1+\epsilon)r)$.
- 2) If all points are outside $\text{Ball}(q, (1+\epsilon)r)$
Then report "No point in $\text{Ball}(q, r)$."
- 3) The intermediate case is immaterial.



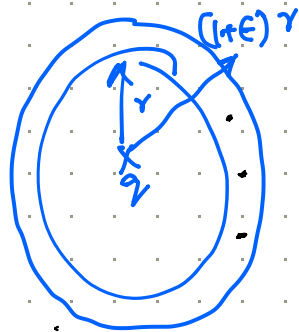
$$\underline{ANN} \cong \underline{APN}$$

$$\underline{APN \leq ANN}:$$

Input to APN: query q , radius r

Directly solve the ANN problem with query q .

Returns p s.t. $\text{dist}(p, q) \leq (1+\epsilon)r$



Two cases: Case 1: $\text{dist}(p, q) \leq (1+\epsilon)r$. Then we return q

Case 2: $\text{dist}(p, q) > (1+\epsilon)r \Rightarrow \text{dist}(\text{nn}(q), q) > r$

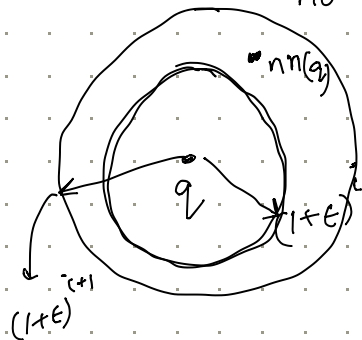
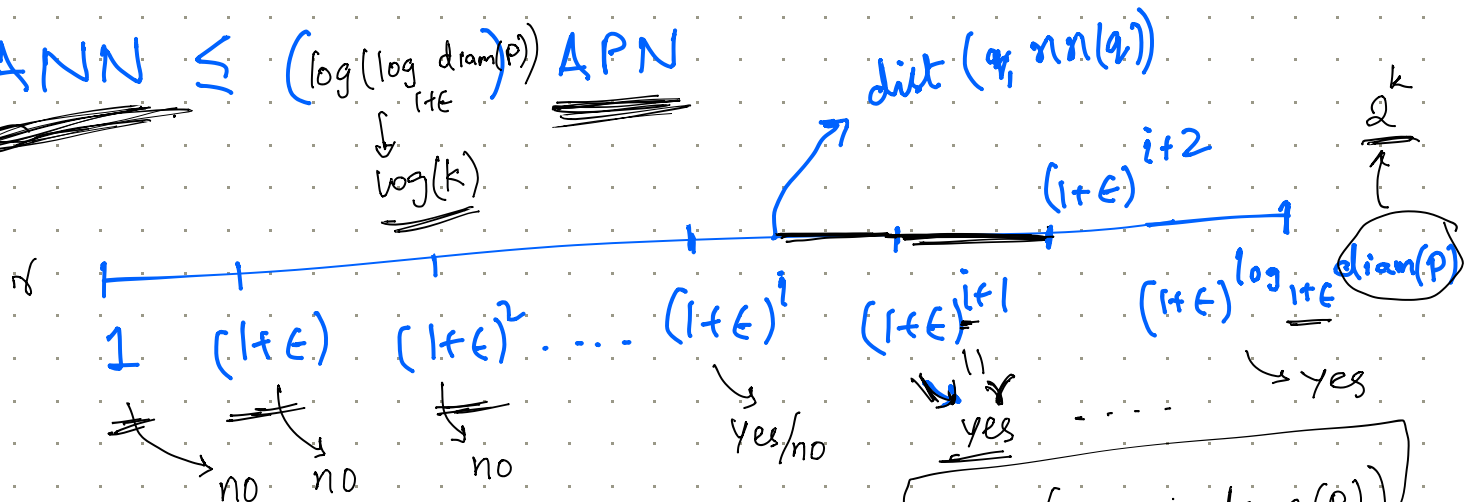
Return "No point inside the ball of radius r ".

$$\underline{\underline{ANN \approx APN}}$$

$$[1, \underline{\underline{\text{diam}(P)}}]$$

$$\underline{\underline{ANN}} \leq \left(\log_{1+\epsilon} (\log \text{diam}(P)) \right) \underline{\underline{APN}}$$

$\downarrow$
 $\log(k)$



$$\log \left(\log_{1+\epsilon} \text{diam}(P) \right)$$

$$\underline{\underline{\text{dist}(P, q)}} \leq (1+\epsilon)^{i+2} = \underline{(1+\epsilon)^2} (1+\epsilon)^i$$

$$\leq \underline{(1+3\epsilon)} \underline{\underline{\text{dist}(\text{nn}(q), q)}}$$

SUMMARY SO FAR

Nearest Neighbor (Hard)

Approximate Nearest Neighbor - ANN

Proximity Query

Approximate Proximity Query - APQ

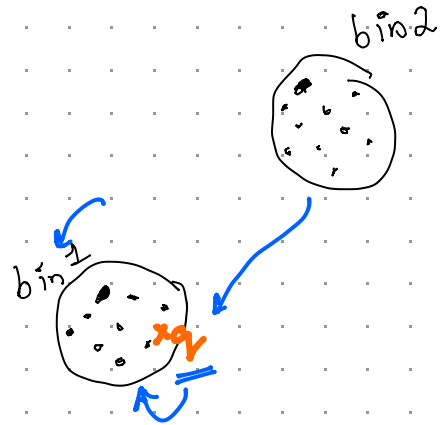
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IDEAL SCENARIO

Store points in P in a hash table such that

Close points $\rightarrow$ Same bin

Far points $\rightarrow$ different bins.



LOCALITY SENSITIVE HASHING (LSH)

A family $\mathcal{F}$ of hash functions is said to be (r, R, α, β) -sensitive if for any $p, q \in \mathcal{U}$, we have the following:

A) If $q \in \text{Ball}(p, r)$, then $\mathbb{P}[f(p) = f(q)] \geq \alpha$

B) If $q \notin \text{Ball}(p, R)$, then $\mathbb{P}[f(p) = f(q)] \leq \beta$

We also require $\alpha > \beta$.

$1 - \beta$

AN LSH FAMILY FOR $(\{0,1\}^d, \|\cdot\|_1)$

The universe is the corners of d-dim unit hypercube.

$f_i(p) = i^{\text{th}}$ coord of p for all $i \in [d]$

Claim: $\mathcal{F} = \{f_1, f_2, \dots, f_d\}$ is (r, R, α, β) -sensitive where

$$\alpha = 1 - \frac{r}{d} \quad \beta = 1 - \frac{R}{d} \quad (\text{for } r \neq R, \alpha > \beta).$$

Proof: (A) $\|p - q\|_1 \leq r$. We chose $f_i \sim \mathcal{F}$. $\mathbb{P}[f_i(p) = f_i(q)] > 1 - \frac{r}{d}$

(B) $\|p - q\|_1 \geq R$. $\mathbb{P}[f_i(p) = f_i(q)] \leq 1 - \frac{R}{d}$

NAIVE APQ^(r) ($\exists \mathcal{F}$ which is $(r, (1+\epsilon)r, \alpha, \beta)$ - LSH)

Preprocessing (The point set is P)

For every $p \in P$, we compute $f(p)$ where $f \sim \mathcal{F}$.

For a query q .

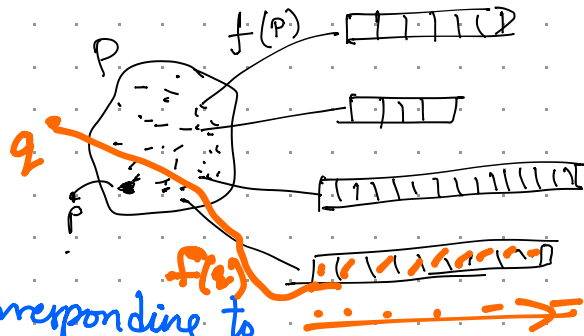
We compute $f(q)$

We search through the list/bucket corresponding to $f(q)$

$f(q)$ for a point p s.t. $\text{dist}(p, q) \leq r(1+\epsilon)$.

If such point p exists, return p .

O/w, say "no point in $\text{Ball}(q, r)$ ".



NAIVE APQ (Analysis)

Suppose query q .

Expec Time to answer query q :- in bucket corresp. to $f(q)$

Once we encounter a point p , s.t. $\text{dist}(p, q) \leq (1+\epsilon)r$.

of points p s.t. $\text{dist}(p, q) > (1+\epsilon)r$:

the prob that p, q match to same bucket
(i.e., $f(p) = f(q)$) is at most β . (β is a constant).

So, # $\leq \underline{\beta(n)}$ [(Constant) n]

NAIVE APQ

$$\& \underline{f(p) = f(q)}$$

If there is p s.t.
 $\text{dist}(p, q) \leq (1+\epsilon)r,$

then we have
a probability
of at least α [$\alpha = \text{constant}$]

$$(f_1, f_2, \dots, f_r) \sim \mathcal{F}$$

We deem that two points p, q
collide iff

$$\nexists p \text{ s.t. } \underline{\text{dist}(p, q) > (1+\epsilon)r},$$
$$\underline{\cong \beta n} \text{ (linear)} \quad \checkmark$$

$$\underline{f} \sim \mathcal{F} \quad \rightarrow \quad \wedge(\mathcal{F}, k)$$
$$(\underline{f_1}, f_2, \dots, f_k) \sim \mathcal{F}$$

We deem that two points $\underline{p \& q}$
"collide" iff

$$f_1(p) = f_1(q) \& f_2(p) = f_2(q) \dots$$

$$\nexists p \text{ s.t. } \text{dist}(p, q) > (1+\epsilon)r \&$$
$$\wedge(\mathcal{F}, k)(p) = \wedge(\mathcal{F}, k)(q) \rightarrow \underline{\beta^k \cdot n} \quad \checkmark$$

$$\underline{f_1(p) = f_1(q)} \quad \underline{\vee f_2(p) = f_2(q) \vee \dots}$$

Then,

Case 1: $\text{dist}(p, q)$ is $\leq (1+\epsilon)r$

$$\begin{aligned} P[V(\mathcal{T}, T)(p) = V(\mathcal{T}, T)(q)] \\ = 1 - (1-\alpha)^T \end{aligned}$$

It can be shown that

$$\# p \text{ s.t. } V(\mathcal{T}, T)(p) = V(\mathcal{T}, T)(q)$$

but $\text{dist}(p, q) \geq (1+\epsilon)r$.

$$\leq \underline{(1 - (1-\beta)^T) n} = (1 - o(1))n$$

Suppose $\text{dist}(p, q) \leq (1+\epsilon)r$,

then

$$f_1(p) = f_1(q) \text{ \& } f_2(p) = f_2(q) \text{ \& } \dots$$

↓

$$P \rightarrow \alpha^k$$

COMBINED APQ

$$\mathcal{H} = \mathcal{V}(\Lambda(\mathbb{F}, k), \mathcal{T}) \longrightarrow k, \mathcal{T} \text{ are unknowns.} \\ \text{will be fixed later.}$$

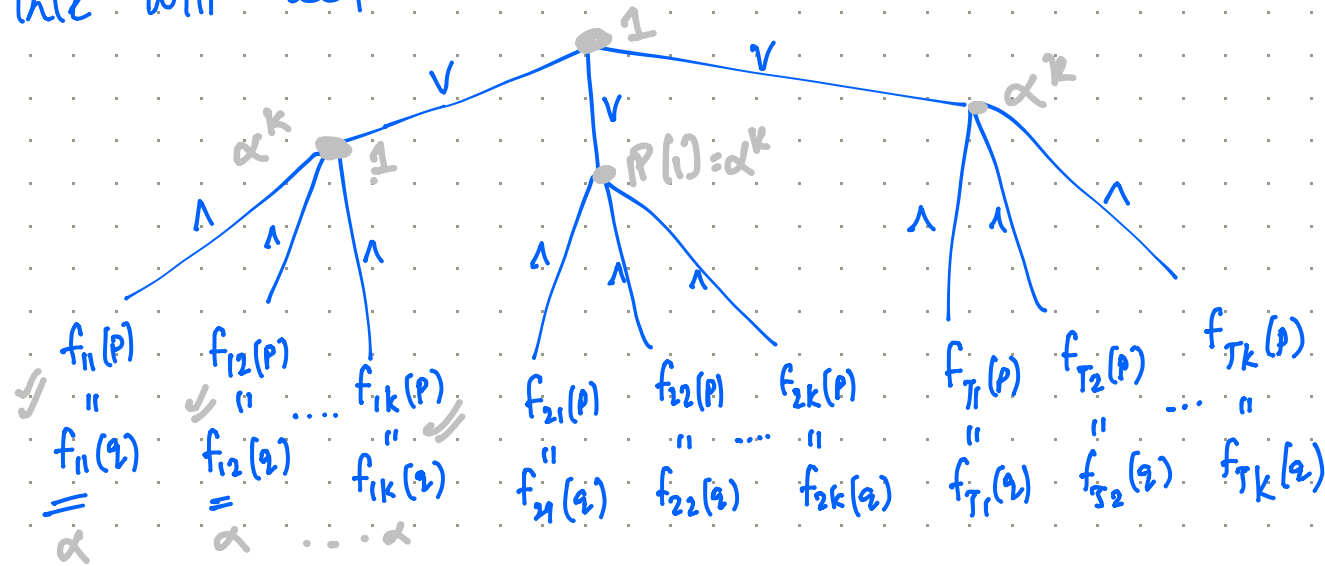
A hash function $h \in \mathcal{H}$, given as input a point P , computes the following matrix.

$$\underline{h(P)} = \begin{bmatrix} f_{11}(P) & f_{12}(P) & \dots & f_{1k}(P) \\ f_{21}(P) & f_{22}(P) & \dots & f_{2k}(P) \\ \vdots & \vdots & \ddots & \vdots \\ f_{\mathcal{T}1}(P) & f_{\mathcal{T}2}(P) & \dots & f_{\mathcal{T}k}(P) \end{bmatrix}$$

each f_{ij} ($i \in [\mathcal{T}]$ and $j \in [k]$) is unif. rand. $\sim \mathbb{F}$.

COMBINED APQ:

We will need a collision criterion to decide when $\underline{h(p)} = \underline{h(q)}$



Lemma: Given an (r, R, α, β) -sensitive LSH $\mathcal{F}$, the family $\mathcal{H} = V(\wedge(\mathcal{F}, k), J)$ is a $(r, R, 1 - (1 - \alpha^k)^J, 1 - (1 - \beta^k)^J)$ -sensitive LSH.

Proof: Suppose $\text{dist}(p, q) \leq r$. We know that $\mathbb{P}_{f \sim \mathcal{F}}[f(p) = f(q)] > \alpha$

$$\text{So, } \mathbb{P}_{h \sim \mathcal{H}}[h(p) = h(q)] \geq 1 - (1 - \alpha^k)^J$$

Similarly, if $\text{dist}(p, q) > R$, we can show that

$$\mathbb{P}_{h \sim \mathcal{H}}[h(q) = h(p)] \leq 1 - (1 - \beta^k)^J$$



Choosing k, T

$$\text{dist}(p, q) \leq \gamma \implies \mathbb{P}[h(p) = h(q)] \geq 1 - (1 - \alpha^k)^T \quad // \quad \alpha > \beta$$

$$\text{dist}(p, q) > (1 + \epsilon)\gamma \implies \mathbb{P}[h(p) = h(q)] \leq \underline{1 - (1 - \beta^k)^T}$$

Impose the condⁿ that $1 - (1 - \alpha^k)^T = 3/4$

$$(1 - \alpha^k)^T = 1/4$$

$$\frac{1}{4} = (1 - \alpha^k)^T \leq \exp(-\alpha^k T) \quad \leftarrow \quad e^{-4} > \underline{\underline{1/4}}$$

It suffices to take $T = \underline{\underline{4/\alpha^k}}$

$$T = 4/\alpha^k$$

Prob that a point in $P \setminus \text{Ball}(q, \underline{(1+\epsilon)\gamma})$ ~~that~~ collides with a given point q is

$$\leq 1 - (1 - \beta^k)^T = (1 - (1 - \beta^k)) \underbrace{\left(1 + \overbrace{(1 - \beta^k)}^{\leq 1} + \overbrace{(1 - \beta^k)^2}^{\leq 1} + \dots \right)}_T$$

$$\leq \beta^k T = \underline{\underline{4 \left(\frac{\beta}{\alpha} \right)^k}}$$

Query time :- $h(q)$ and then ~~to~~ go thru the Bucket B which contains all points p s.t. $h(q) = h(p)$.

If we find p s.t. $\text{dist}(p, q) \leq (1 + \epsilon)\gamma$,

If the num of points p s.t. $\text{dist}(p, q) > (1 + \epsilon)\gamma$

Expec num of P s.t $\text{dist}(P, q) > (1+\epsilon) \gamma$ and $h(p) = h(q)$
is given by $4n \left(\frac{\beta}{\alpha}\right)^k$.

and to compute $h(q)$ itself, we need $\mathcal{I}k$ time.

$$\text{Query time} \propto \left(\mathcal{I}k + 4n \left(\frac{\beta}{\alpha}\right)^k \right) d //$$

By Take $\boxed{\mathcal{I}k = 4n \left(\frac{\beta}{\alpha}\right)^k}$

$$= n \beta^k \mathcal{I}$$

$\Rightarrow$

$$\boxed{k = n \beta^k}$$

If we choose

$$\boxed{k = \log_{(\frac{1}{\beta})} n} //$$

$$\left(\frac{1}{\beta}\right)^k = n$$

$$\boxed{n \beta^k = 1}$$

$$\text{Query time} = d \left(\tau k + 4n \left(\frac{1}{\alpha} \right)^k \right)$$

$$\leq d \left(\tau k + 4 \frac{(k)}{\alpha^k} \right)$$

$$= \underline{\underline{2d\tau k}} \approx \underline{\underline{O(d\tau k)}} \checkmark$$

$$k = \frac{\ln n}{\ln(1/\beta)}$$

$$\tau = \frac{4}{\alpha^k}$$

$$= \frac{4}{\alpha^{\frac{\ln n}{\ln(1/\beta)}}} = \frac{4}{\left(\alpha^{\ln n} \right)^{1/\ln(1/\beta)}} \\ = \frac{4}{\left(n^{\ln(\alpha)} \right)^{1/\ln(1/\beta)}}$$

$$\alpha = 1 - \frac{r}{d}, \quad \beta = 1 - \frac{(1+\epsilon)r}{d}$$

$$= 4 n \frac{\ln(\frac{1}{\alpha})}{\ln(1/\beta)}$$

$$= \underline{T}$$

$$\text{Query time} \approx 2d \left(4 n \frac{\ln(1/\alpha)}{\ln(1/\beta)} \cdot \frac{\ln(h)}{\ln(1/\beta)} \right)$$

$$\approx O \left(n \frac{\ln(1/\alpha)\epsilon}{\ln(1/\beta)} \ln(n) d \right)$$

Prop: For $x \in (0,1)$ & $t \geq 1$ s.t. $1-tx > 0$, we have

$$\frac{\ln\left(\frac{1}{1-x}\right)}{\ln\left(\frac{1}{1-tx}\right)} \leq \frac{1}{t}$$

A) $g(x) = \frac{(1-tx) - (1-x)^t}{1}$

$$g'(x) = -t + t(1-x)^{t-1}$$

$$= t \left((1-x)^{t-1} - 1 \right)$$

$$\leq 0$$

$$g(0) = (1-0) - (1-0)^t = 0.$$

$$g(x) \leq 0 \quad \text{for all } x \in (0,1).$$

$$(1-tx) \leq (1-x)^t$$

$$\ln(1-tx) \leq t \ln(1-x)$$

$$\ln\left(\frac{1}{1-tx}\right) \geq t \ln\left(\frac{1}{1-x}\right)$$

$$\Rightarrow \frac{\ln\left(\frac{1}{1-x}\right)}{\ln\left(\frac{1}{1-tx}\right)} \leq \frac{1}{t}.$$

$$\underline{\text{Query time}} = O \left(n^{\frac{\ln(\frac{1}{1-r/d})}{\ln(\frac{r}{(1+\epsilon)r/d})}} (\ln n) d \right)$$

$$\leq O \left(\underbrace{n^{\frac{1}{1+\epsilon}}}_{n^{1/3}} (\ln n) d \right)$$

Preprocessing space :-

$$(n)(d)(\underline{Tk}) = \underbrace{O \left(n^{1+\frac{1}{1+\epsilon}} (\ln n) d \right)}_{n^{5/2}}$$

Main Thm : (Assuming $(\mathcal{K}, \text{dist})$ has (r, R, α, β) -LSH family)

(APQ) It is possible to preprocess a set of n points in

$\mathbb{R}^d$ such that

Preprocessing space is $O\left(n^{1 + \frac{\ln(1/\alpha)}{\ln(1/\beta)}} (\ln n) d\right)$

Query time is $O\left(n^{\frac{\ln(1/\alpha)}{\ln(1/\beta)}} (\ln n) d\right)$

Each query is successful with prob at least $\frac{3}{4}$.

LSH on Euclidean Metric ($\mathbb{R}^d$, dist)

Lemma:- Let $X_1, X_2, \dots, X_d$ be normally distributed independent random variables. Let $v_1, \dots, v_d \in \mathbb{R}$. Then the variable $Y = v_1 X_1 + v_2 X_2 + \dots + v_d X_d$ will be distributed as $\mathcal{N}(0, \sigma^2 = \|\vec{v}\|^2)$ where $\vec{v} = (v_1, v_2, \dots, v_d)$

Proof:- $\phi_x(t) = \mathbb{E} [e^{itx}]$

It can be shown that for $\mathcal{N}(0, \sigma^2)$, the

$$\phi_{N(0, \sigma^2)}(t) = \exp\left(-\frac{\sigma^2 t^2}{2}\right)$$

$$Y = v_1 X_1 + \dots + v_d X_d$$

$$\phi_Y(t) = \mathbb{E} \left[e^{it(v_1 X_1 + v_2 X_2 + \dots + v_d X_d)} \right]$$

$$= \prod_{j=1}^d \mathbb{E} \left[e^{it(v_j X_j)} \right]$$

$$= \prod_{j=1}^d \exp\left(-\frac{v_j^2 t^2}{2}\right)$$

$$\phi_Y(t) = \exp\left(-\left(v_1^2 + v_2^2 + \dots + v_d^2\right) \frac{t^2}{2}\right)$$

$$\gamma \sim \mathcal{N}(0, \sigma^2 = \|\mathbf{v}\|^2).$$



LSH for Euclidean Metric:

Fix a vector $\vec{v}$. Then, let's define for a point $p \in \mathbb{R}^d$

$$h(p) = \left\lfloor \frac{\langle p, \vec{v} \rangle + t}{x} \right\rfloor \quad \text{where } t \sim \text{Unif}[0, x] \text{ and}$$

x is a parameter which we will choose shortly.

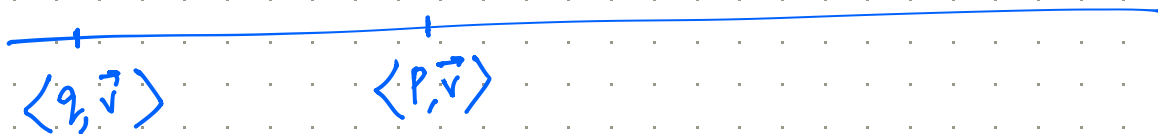
$$\mathcal{H} = \left\{ h : \vec{v} \text{ is a vector } \sim \underline{\mathcal{N}(0, 1)^d} \right\}$$

We need to choose α s.t. $\mathcal{H}$ is LSH.

Fix $p, q, \vec{v}, \alpha$. Let $\|p - q\| =: \eta$ and

$$|\langle p, \vec{v} \rangle - \langle q, \vec{v} \rangle| =: \beta.$$

$$\langle p, \vec{v} \rangle - \langle q, \vec{v} \rangle = \beta \quad (\text{Assume})$$



Fix some $p, q, \vec{v}, \alpha$.

$$\mathbb{P} [h(p) = h(q)]$$

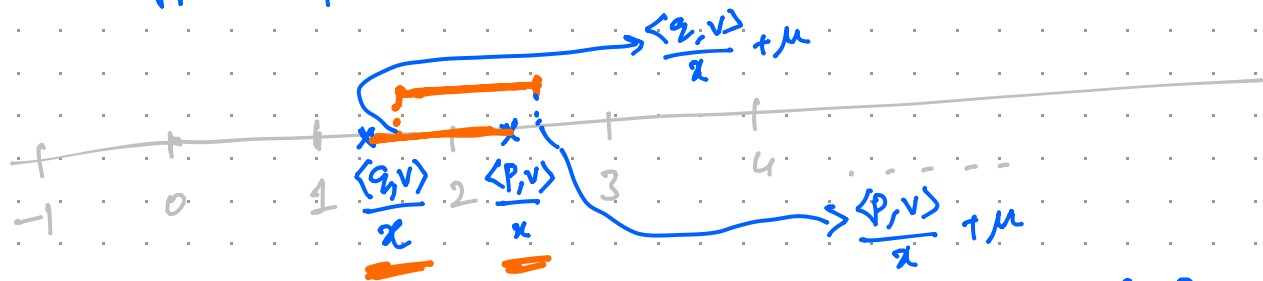
$$h(p) = h(q) \iff \left\lfloor \frac{\langle p, \vec{v} \rangle + t}{\alpha} \right\rfloor = \left\lfloor \frac{\langle q, \vec{v} \rangle + t}{\alpha} \right\rfloor$$

Claim: If $\beta = \langle p, \vec{v} \rangle - \langle q, \vec{v} \rangle > \alpha$, then $h(p) \neq h(q)$.

Pf:
$$\begin{aligned} h(p) &= \left\lfloor \frac{\langle q, \vec{v} \rangle + \beta + t}{\alpha} \right\rfloor = \left\lfloor \frac{\beta}{\alpha} + \frac{\langle q, \vec{v} \rangle + t}{\alpha} \right\rfloor \\ &> \left\lfloor \frac{\langle q, \vec{v} \rangle + t}{\alpha} \right\rfloor = h(q) \end{aligned}$$

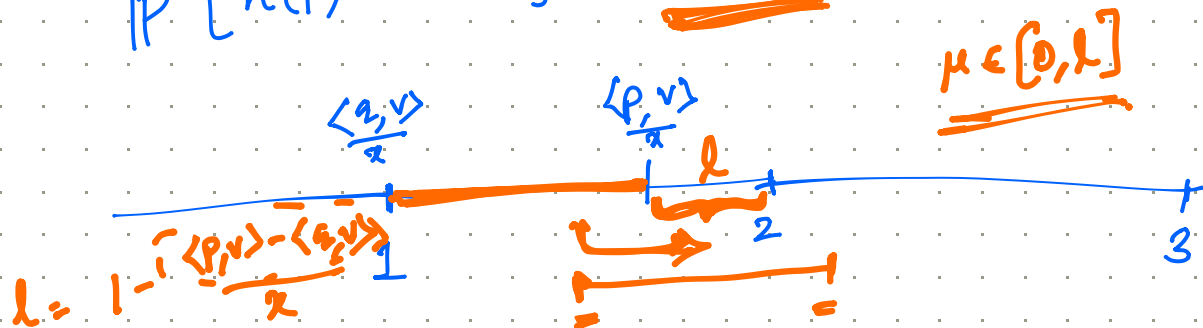


Suppose $\beta = \langle p, v \rangle - \langle q, v \rangle \leq x$.



$$h(p) = \left\lfloor \frac{\langle p, v \rangle}{x} + \mu \right\rfloor \quad \text{where } \mu \sim \text{Unif}[0, 1]$$

$$\mathbb{P}[h(p) = h(q)] = \underline{1 - \frac{\beta}{x}} \quad \left(\beta = \langle p, v \rangle - \langle q, v \rangle \right)$$



$$P[h(p) = h(q)] = \int_0^{\beta} P\left[\underbrace{|\langle p, v \rangle - \langle q, v \rangle| = \beta}_{\text{...}}\right] \underbrace{\left(1 - \frac{\beta}{\alpha}\right) d\beta}$$

$$\langle p, v \rangle - \langle q, v \rangle = \langle \underline{p - q}, v \rangle \sim \mathcal{N}(0, \|p - q\|^2)$$

Say $\eta = \underline{\|p - q\|}$.

$$\begin{aligned} P\left[|\langle p, v \rangle - \langle q, v \rangle| = \beta\right] &= P\left[\langle p - q, v \rangle = \beta \text{ or } \langle q - p, v \rangle = \beta\right] \\ &= 2 \, P\left[\underline{\langle p - q, v \rangle = \beta}\right] \end{aligned}$$

$$\underline{P \left[\langle p-q, v \rangle = \beta \right]_{\beta}^x \rightarrow \text{the pdf of } \mathcal{N}(0, \eta^2) \text{ at } \beta}$$

$$2 \left[\text{pdf of } \mathcal{N}(0, \eta^2) \right]_{\text{at } \beta}^x = \frac{2}{\sqrt{2\pi} \eta} \exp\left(-\frac{\beta^2}{2\eta^2}\right)$$

$$P \left[h(p) = h(q) \right] = \int_0^x \underbrace{\frac{2}{\sqrt{2\pi} \eta} \exp\left(-\frac{\beta^2}{2\eta^2}\right) \left(1 - \frac{\beta}{x}\right)}_{\text{" } \|p-q\|} d\beta \quad p(\eta, x)$$

$$\tilde{O} \left(n \frac{\ln(1/\alpha)}{\ln(1/\beta)} \right) \quad \text{where} \quad \alpha \leq P[h(p) = h(q)] \text{ given } \|p - q\| \leq \epsilon$$

$$\beta \geq P[h(p) = h(q)] \text{ given } \|p - q\| \geq r(1 + \epsilon)$$

Choose α s.t

$$\frac{\ln \left(\frac{1}{P(n, \alpha)} \right)}{\ln \left(\frac{1}{P(n(1+\epsilon), \alpha)} \right)}$$

is minimized

$$\ln \left(\frac{1}{P(n(1+\epsilon), \alpha)} \right)$$

Datar, Immorlica, Indyk, Mirrokni (2004)

Using computations, they show that it is possible to
choose param α s.t

$$\frac{\ln \left(\frac{1}{p(\eta, \alpha)} \right)}{\ln \left(\frac{1}{p(\eta(1+\epsilon), \alpha)} \right)} \leq \frac{1}{1+\epsilon} \quad \text{for all } \eta > 0$$

→ APQ!

Query time $\rightarrow \tilde{O} \left(d n^{\frac{1}{1+\epsilon}} \right)$

Preprocessing space $\rightarrow \tilde{O} \left(d \left(n^{1+\frac{1}{1+\epsilon}} \right) \right)$.