Lecture Notes-2

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$$\begin{split} \overrightarrow{X} &= (X_1, X_2, ..., X_d) |X|^2 \\ &= \sum_{i=1}^d X_i^2 E[|X|^2] \\ &= E[X_1^2] + E[X_2^2] + ... + E[X_d^2] \\ &= |X| \ O(\sqrt{d}) \end{split}$$

1 Strong Law of Large Numbers (SLLN)

n points P[—-O(\sqrt{d})— $\geq \epsilon$] $\leq \frac{Var(X)}{n\epsilon^2}$

If we chose two points among n points, what kind of configuration will they have on average?

$$\begin{aligned} |Y - Z|^2 &= \sum_i |Y_i - Z_i|^2 E[|Y - Z|^2] \\ &= E[\sum_i |Y_i - Z_i|^2] \\ &= E(\sum_i (Y_i^2 + Y_i^2 + 2Y_i \times Z_i) \\ &= d + d = 2d \\ &\quad E[Y^2] = d \\ &\quad E[Z^2] = d \end{aligned}$$

2 Volume/Area of Unit Ball in d-dimensions

 $\frac{V(d)}{A(d)}$ represents the unit ball. In case of having a ball with radius r, the ratio becomes $\frac{V(r)}{A(r)}$

$$V(d) = \int_{x_1=-1}^{x_1=1} \dots \int_{-\sqrt{1-x_1^2-\dots-x_{d-2}^2}}^{\sqrt{1-x_1^2-\dots-x_{d-2}^2}} \int_{-\sqrt{1-x_1^2-\dots-x_{d-1}^2}}^{\sqrt{1-x_1^2-\dots-x_{d-1}^2}} dx_d dx_{d-1} \dots dx_1$$
 (1)

When d=2, $d\Omega \gamma^{d-1} d\gamma$

$$\int_0^{2\pi} d\Omega \int_0^1 \gamma d\gamma = \pi \tag{2}$$

$$V(d) = \int_{S_d} d\Omega \int_0^1 \gamma^{d-1} d\gamma = \frac{A(d)}{d}$$
 (3)

$$I(d) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} e^{-(x_1^2 + x_2^2 + \dots + x_d^2)} dx_1 dx_2 \dots dx_d$$

$$= \left(\int_{-\infty}^{\infty} e^{-x^2} dx \right)^d$$

$$= \sqrt{\pi}^d$$

$$= \pi^{d/2}$$

$$(4)$$

2.1 Spherical Coordinates

$$\begin{split} I(d) &= \int_{S_d} d\Omega \int_0^\infty e^{-\gamma^2} \gamma^{d-1} d\gamma \\ &= A(d) \int_0^{-\infty} e^{-\gamma^2} \gamma^{d-1} d\gamma \\ &= A(d) \int_0^\infty e^{-t} t^{\frac{d-1}{2}} \frac{t^{-\frac{1}{2}}}{2} dt \\ &= A(d) \int_0^\infty e^{-t} t^{\frac{d}{2}-1} \\ &= \frac{1}{2} \Omega(\frac{d}{2}) A(d) \end{split}$$

Then,

$$A(d) = \frac{\pi^{\frac{d}{2}}}{\frac{\Gamma(\frac{d}{2})}{2}} \tag{5}$$

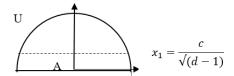
and

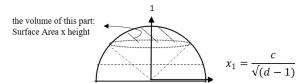
$$V(d) = \frac{\pi^{\frac{d}{2}}}{\frac{d \times \Gamma(\frac{d}{2})}{2}} \tag{6}$$

Theorem 1. Given $c \ge 1$ and $d \ge 3$, at least $1 - \frac{2e^{\frac{-c^2}{2}}}{c}$ fraction of the volume of d-dimensional unit ball has $|X_1| \le \frac{c}{\sqrt{d-1}}$

$$\frac{Vol(U/A)}{Vol(U)} \leq \frac{\text{upper bound of Vol(U/A)}}{\text{lower bound of Vol(U)}} \approx \frac{2e^{\frac{-c^2}{2}}}{c}$$

How to find an upper bound on Vol(A)?





$$Vol(U/A) = V(d-1) \int_{\frac{c}{\sqrt{d-1}}}^{1} (1 - x_1^2)^{\frac{d-1}{2}} dx_1$$

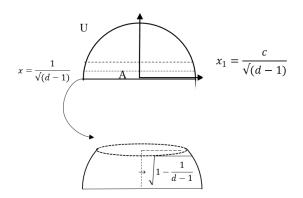
$$\leq V(d-1) \int_{\frac{c}{\sqrt{d-1}}}^{1} \frac{x_1 \sqrt{d-1}}{c} (1 - x_1^2)^{\frac{d-1}{2}} dx_1$$

$$\leq \frac{(d-1)\sqrt{d-1}}{c} \int_{\frac{c}{\sqrt{d-1}}}^{\infty} x_1 e^{\frac{-(d-1)x_1^2}{2}} dx_1$$

$$= \frac{e^{\frac{-c^2}{2}}}{d-1}$$

$$Vol(U/A) \leq \frac{Vol(d-1)}{c\sqrt{d-1}} e^{\frac{-c^2}{2}}$$

$$(7)$$



Then the volume is lower bound on U

$$\left(1 - \frac{1}{d-1}\right)^{\frac{d-1}{2}} V(d-1) \frac{1}{\sqrt{d-1}} \tag{8}$$

Consider the formula $(1 - \mathbf{a}x) \le (1 - x)^{\mathbf{a}}$, $\mathbf{a} \ge 1$

Replace $\mathbf{a}=\frac{d-1}{2}$ and $\mathbf{x} \mathbf{=} \frac{1}{d-1} \mathbf{,}$ then (8) becomes

$$1 - \left(\frac{d-1}{2}\right) \times \left(\frac{1}{d-1}\right) = \frac{1}{2} \tag{9}$$

Theorem 2. N points $X_1,...,X_N$ from d-dimensional unit ball with probability $1-O(\frac{1}{n})$

1. •
$$|X_i| \ge 1 - \frac{2lnN}{d}$$
, $P[|X_i| < 1 - \frac{2lnN}{d}] \le e^{-d\frac{2lnN}{d}} \le e^{-\epsilon d}$

2. •
$$|X_i - X_j| \le \frac{\sqrt{6lnN}}{\sqrt{d-1}}$$
, $P[|X_i| > \frac{c}{\sqrt{d-1}}] \le \frac{2e^{-\frac{c^2}{2}}}{2}$

Take N points, among all these points, there are $\binom{N}{2}$ pairs X_i- "North $X_j\to X_i$ $(X_j$ projected on $X_i)$

Projection larger than
$$\frac{\sqrt{6lnN}}{d-1}$$
 has order $O\left(\frac{1}{N^3}\right)$ and $O\left(\binom{N}{2}\frac{1}{N^3}\right)=O(\frac{1}{N})$

3 Chernoff Bounds

(Sums of Poisson Trials,Extension of Tcheybshev Bounds) X is a random variable and t is given. Then Moment Generating Function $E[e^{tX}]=M_X(t)$ and $M_X^n(0)=E[X^n]$

From Markov Inequality,

t>0,
$$P[X \ge \mathbf{a}] = P[e^{tX} \ge e^{ta}] \le min_{t>0} \frac{E[e^{tX}]}{e^{t\mathbf{a}}}$$
 t<0, $P[X \le \mathbf{a}] \le min_{t<0} \frac{E[e^{tX}]}{e^{t\mathbf{a}}}$

3.1 Poisson Trials

$$X_1,X_2,...,X_N \text{ are independent and } X_i = \begin{cases} 1 \text{wp} p_i \\ 0 w p (1-p_i) \end{cases}$$

$$\mathbf{X} = \sum_{i=1}^N X_i, \ \mu = E[X] = \sum_{i=1}^N b_i$$

$$M_X(t) = E[e^{tX}] = \left(E[e^{tX_i}]\right)^N \tag{10}$$

$$M_{X_i}(t) = E[e^{tX_i}]$$

$$= p_i e^t + (1 - p_i)e^0$$

$$= \underbrace{1 + p_i(e^t - 1) \le e^{p_i(e^t - 1)}}_{(1+x) \le e^x}$$
(11)

Then, equation (10)

$$E[e^{tX_i}]^N \le \prod_{i=1}^N e^{p_i(e^t - 1)}$$

$$= e^{\sum_{i=1}^N p_i(e^t - 1)}$$

$$= e^{\mu(e^t - 1)}$$
(12)

Result:

$$\delta > 0, P[X \ge (1+\delta)\mu] \le \left[\frac{e^{\delta}}{(1+\delta)^{1+\delta}}\right]$$
(13)